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Investigations Into the Analysis and Modeling of the TNF α -Mediated NF- κ B-Signaling Pathway

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In this study, we propose a system-theoretic approach to the analysis and quantitative modeling of the TNF α -mediated NF- κ B-signaling pathway. Tumor necrosis factor α (TNF α) is a potent proinflammatory cytokine that plays an important role in immunity and inflammation, in the control of cell proliferation, differentiation, and apoptosis. To date, there have been numerous approaches to model cellular dynamics. The most prominent uses ordinary differential equations (ODEs) to describe biochemical reactions. This approach can provide us with mathematically well-founded and tractable interpretations regarding pathways, especially those best described by enzyme reactions. This work first introduces a graphical method to intuitively represent the TNF α -mediated NF- κ B-signaling pathway and then utilizes ODEs to quantitatively model the pathway. The simulation study shows qualitative validation of the proposed model compared with experimental results for this pathway. The proposed system-theoretic approach is expected to be further applicable to predict the signaling behavior of NF- κ B in a quantitative manner for any variation of the ligand, TNF α .

Tumor necrosis factor α (TNF α) is a potent proinflammatory cytokine that plays an important role in immunity and inflammation and in the control of cell proliferation, differentiation, and apoptosis. Binding of TNF α to its two receptors, TNFR1 and TNFR2, results in recruitment of signal transducers that activate at least three distinct effectors. Through complicated signaling cascades and networks, these effectors lead to the activation of caspases and two transcription factors, AP-1 and NF- κ B (Francis et al. 2000; Swaroop and David 2001; Veronique and Michael 2001; Michael and Anning 2002). This study is to propose a system-theoretic approach for the analysis and quantitative modeling of the TNF α -mediated NF- κ B-signaling pathway.

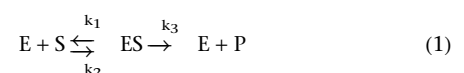
We can find numerous approaches in the literature for the modeling of cellular dynamics (Robert 1997; Upinder and Ravi 1999; Anand and Douglas 2000; Douglas 2000; Jeff et al. 2001; Robert and Tom 2001; Tyson et al. 2001; Wolkenhauer 2001). The most prominent approach uses ordinary differential equations (ODEs) to model biochemical reactions. This can provide us with mathematically well-founded and tractable interpretations for biological pathways, especially those best described by enzyme reactions. Recently, Hoffmann et al. (2002) presented a remarkable computational model that describes the temporal control of NF- κ B activation by the coordinated degradation and synthesis of I κ B proteins. The focus of the model was the central-signaling module composed of IKK, three mammalian I κ B isoforms, and NF- κ B to address the role of different I κ B isoforms in controlling the dynamic response of the pathway to transient and persistent stimulation. The model was developed in a heuristic way considering a simple differential equations approximation of a biochemical system with negative feedback and provided valuable conclusions regarding the respective functions of I κ B isoforms, the bimodal signal-processing characteristics, and the newly revealed mechanism for specificity in gene expression. In this work, we first focus on developing a general modeling

framework for signal-transduction pathways on the basis of a system-theoretic approach from complemented enzyme-reaction kinetics, and then apply the developed methodology to the TNF α -mediated NF- κ B-signaling pathway. The model also captures the dynamics regarding generation and degradation of the enzyme and provides a basis for sensitivity analysis of the pathway with respect to the variation of initial concentrations (TNF α) in a three-dimensional graphical representation. Specifically, this study first introduces a graphical method to intuitively represent the TNF α -mediated NF- κ B-signaling pathway and then utilizes ODEs to quantitatively model the pathway dynamics. For this purpose, we complement the enzyme kinetics to incorporate the restoring process into the steady-state concentration of substrates. We then model the pathway step-by-step, on the basis of these complemented enzyme kinetics. By integrating individual ODE models into a whole, the quantitative model of the signaling pathway is established. The simulation study shows qualitative validation of the proposed model on the basis of current experimental results. Each parameter value for the computer simulation is derived via inference on the basis of the experimental reaction time.

The work is organized as follows. Section I briefly introduces some preliminaries on mathematical modeling of reaction kinetics are described above. Section II describes the TNF α -mediated NF- κ B-signaling pathway. Section III presents the proposed graphical and mathematical modeling of the TNF α -mediated NF- κ B-signaling pathway on the basis of ODEs. Section IV illustrates simulation studies. Finally, conclusions are made in Section V.

METHODS AND RESULTS

Preliminaries



Equation 1 is a very well-known simple model for enzyme kinetics. It is often used to account for typical kinetic properties of various kinds of enzymes. In equation 1, E is the concentration of an enzyme that combines with a substrate S to form an enzyme-substrate complex ES with a rate constant k_1 . The complex ES

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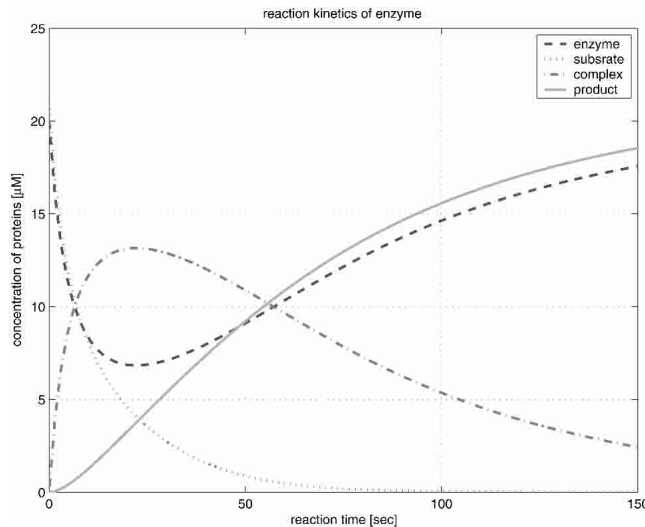


Figure 1 Reaction profile of the basic enzyme kinetics.

holds two possible outcomes in the next step. It can be dissociated into E and S with a rate constant k_2 , or it can further proceed to form a product P with a rate constant k_3 . It is assumed that none of the products reverts to the initial substrate. It is necessary to express the relations between the rate of catalysis and the change of concentration for the substrate, the enzyme, the complex, and the product. These relations are usually represented by the following set of ODEs. The reaction kinetics we develop in the remainder of this work are based on these ODEs.

$$\begin{aligned} \frac{dE(t)}{dt} &= k_1 \cdot E(t) \cdot S(t) + (k_2 + k_3) \cdot ES(t) \\ \frac{dS(t)}{dt} &= k_1 \cdot E(t) \cdot S(t) + k_2 \cdot ES(t) \\ \frac{dES(t)}{dt} &= k_1 \cdot E(t) \cdot S(t) - (k_2 + k_3) \cdot ES(t) \\ \frac{dP(t)}{dt} &= k_3 \cdot ES(t) \end{aligned} \quad (2)$$

Figure 1 illustrates the reaction profile of the basic enzyme kinetics in equation 1 evolving in time. It shows that the formation of an enzyme-substrate complex is the first step in enzyme catalysis. Initially, both the concentration of the substrate and the enzyme decrease until after ~25 sec; the concentration of the

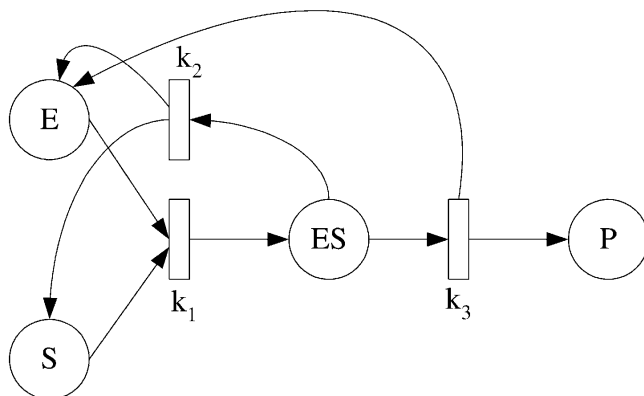


Figure 2 Graphical representation of the reaction kinetics.

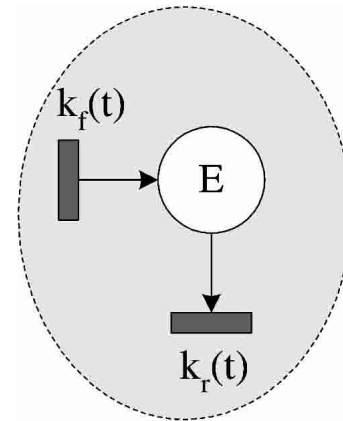


Figure 3 The complemented graphical model of an enzyme.

enzyme increases before settling to its steady state. On the other hand, the concentration of the complex reaches its maximum at ~25 sec, and then decreases exponentially. The curve for the concentration of the product increases and reaches its steady state long after these transitional changes. In reaction kinetics, the change of each concentration implies a signal transfer, passing information to other agonists. The trail of this signal transfer constitutes a signaling pathway. Some signals in a signaling pathway can be merged and amplified or reduced due to interference among signals.

A popular way to intuitively describe a signaling pathway utilizes a graphical representation, akin to diagrams familiar to, or popular with, biologists. These cartoons are usually insufficiently descriptive to explain molecular dynamics quantitatively, but they provide an intuition of the overall dynamics and information processing in cellular systems. Here, we propose both a graphical method and a mathematical method on the basis of ODEs to investigate the analysis and modeling of TNF α -mediated NF- κ B-signaling pathway.

The graphical tool is based on a bipartite-directed multi-graph, in which structure of the bipartite graph consists of two types of nodes and directed arcs; a circle represents a state for the concentration of a protein, a bar represents a rate of reaction, and directed arcs (arrows) connect the circles and the bar. The signal flow in a signaling pathway can then be qualitatively described by this graphical tool.

Equation 2 shows the differential equation model of the reaction kinetics outlined in equation 1, and Figure 2 illustrates the graphical representation for the kinetics in equations 1 and 2.

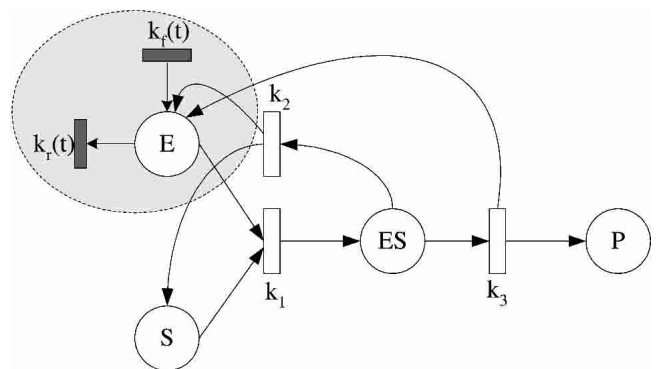


Figure 4 Graphical representation of the complemented model of the reaction kinetics.

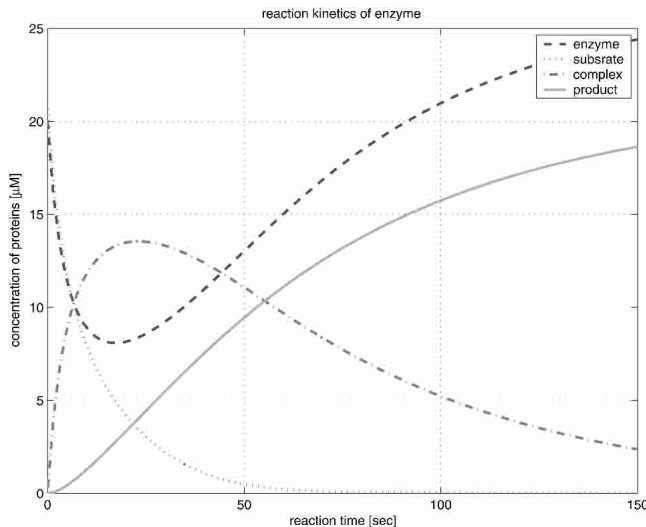


Figure 5 Simulation results of the complemented reaction kinetics.

In Figure 2, the bar represents the rate of formation and breakdown of the complex or the product.

Because the dynamics regarding generation and degradation of the enzyme is not included in the conventional representation of equations 1 and 2, we complement the enzyme kinetic model 2 by considering these dynamics. The main idea is based on the observation that the signal-transduction system usually behaves as a slowly time-varying nonlinear system during the reaction period. Figure 3 illustrates the complemented model of an enzyme (or protein) including the generation and degradation, whose role is to maintain the steady-state concentration of the enzyme (or protein). In Figure 3, the two bars for $k_f(t)$ and $k_r(t)$ represent the rates of generation and degradation, respectively. If we consider the fact that the steady state of enzymes or proteins depends upon the local environment in the cell, it is reasonable to model the rate as a function of time.

Hence, the concentration of an enzyme (or protein) can be modeled as follows.

$$\frac{dE(t)}{dt} = k_f(t) - k_r(t) \cdot E(t) \quad (3)$$

On the basis of assumption of slowly time-varying processes, we can approximate the rates for the generation and degradation as constant during the reaction period, which means that the steady-state concentration of the enzyme (or protein) is independent of time. Thus,

$$k_f(t) \equiv k_f(t_0), \quad k_r(t) \equiv k_r(t_0). \quad (4)$$

By substituting equation 4 into equation 3 and solving the resultant differential equation, we have

$$E(t) = \frac{k_f(t_0)}{k_r(t_0)} - \left(\frac{k_f(t_0)}{k_r(t_0)} - E_0 \right) \cdot e^{-k_r(t_0) \cdot t}. \quad (5)$$

Taking the limit in equation 5, we obtain the following steady-state value for $E(t)$,

$$\lim_{t \rightarrow \infty} E(t) = \frac{k_f(t_0)}{k_r(t_0)}. \quad (6)$$

Finally, we have obtained the following complemented reaction kinetics model on the basis of ODEs, including the rate of generation and degradation for an enzyme (or protein):

$$\begin{aligned} \frac{dE(t)}{dt} &= \left[\frac{k_f(t) - k_r(t) \cdot E(t)}{k_r(t_0)} \right] - k_1 \cdot E(t) \cdot S(t) + (k_2 + k_3) \cdot ES(t) \\ \frac{dS(t)}{dt} &= -k_1 \cdot E(t) \cdot S(t) + k_2 \cdot ES(t) \\ \frac{dES(t)}{dt} &= k_1 \cdot E(t) \cdot S(t) - (k_2 + k_3) \cdot ES(t) \\ \frac{dP(t)}{dt} &= k_3 \cdot ES(t) \end{aligned} \quad (7)$$

In the extreme, if $k_f(t_0) \gg k_r(t_0)$, then $E(t) \approx 0$. This implies a case when the signal flow is blocked, as there is no more signaling protein involved in its signaling pathway. The case depends wholly upon the cell environment.

Figures 4 and 5 provide a graphical representation of the complemented model for the reaction kinetics and the corresponding simulation results for equation 7.

Figure 5 shows a similar behavior to Figure 1, on the basis of equation 2. Note that the curve for the enzyme returns to its steady state more quickly in Figure 5. If we reduce the rate of

Table 1. Definition of Variables and Nominal Values for Signaling Proteins at Steady State

Dependent variables	Acronym	Symbol	Concentration [μM]	Independent variables	[μMsec ⁻¹]/[sec ⁻¹]
TNF Necrosis Factor	TNF	m1	20	—	—
TNF Receptor 1	TNFR1	m2	25	k_{2f}/k_{2r}	0.139/0.00556
TNFR1-complex	—	m3	0	—	—
TNFR1-associated death domain protein	TRADD	m4	25	k_{4f}/k_{4r}	0.139/0.00556
TRADD-complex	—	m5	0	—	—
Fas-associated death domain	FADD	m6	0	k_{6f}/k_{6r}	0.139/0.00556
FADD-complex	—	m9	0	—	—
Receptor-interacting protein	RIP	m7	25	k_{7f}/k_{7r}	0.139/0.00556
RIP-complex	—	m10	0	—	—
TNF-Receptor-associated factor 2	TRAF2	m8	12	k_{8f}/k_{8r}	0.139/0.00556
TRAF2-complex	—	m11	0	—	—
I κ B kinase	IKK	m9	25	k_{9f}/k_{9r}	0.139/0.00556
NF- κ B Inhibitor/Nuclear factor κ B	I κ B/NF- κ B	m14	25	k_{14f}/k_{14r}	0.139/0.00556
IKK-complex	—	m12	0	—	—
NF- κ B Inhibitor	I κ B	m16	0	—	—
Nuclear factor κ B	NF- κ B	m17	0	—	—

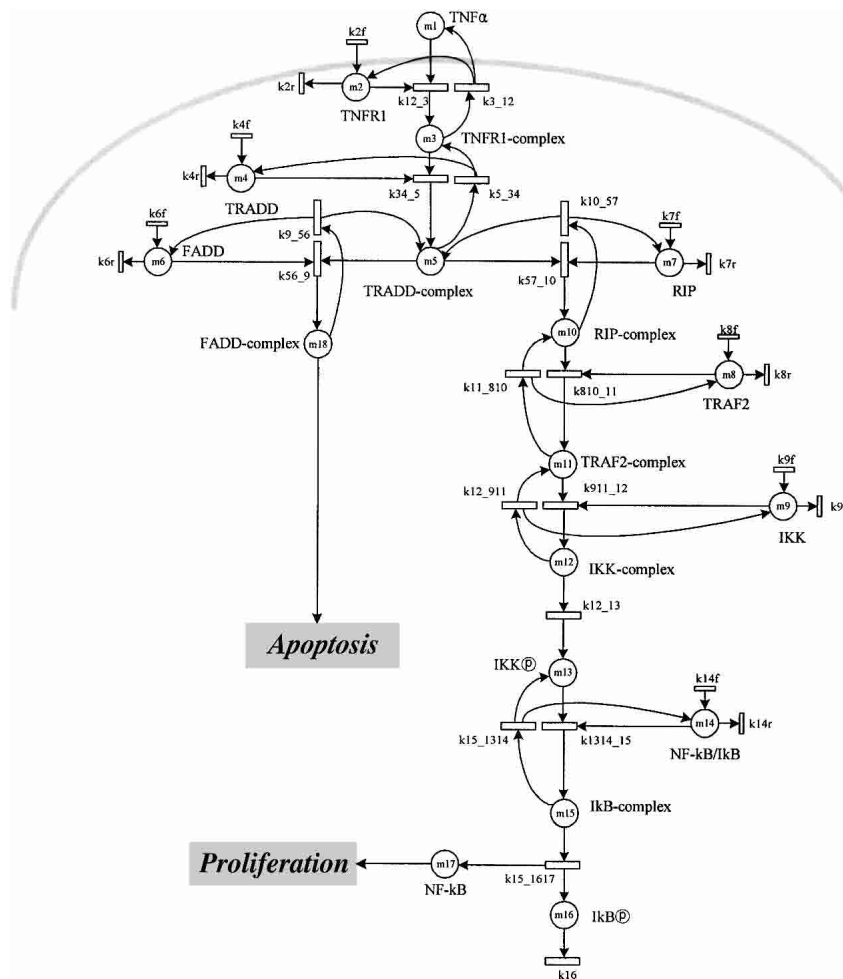


Figure 6 Graphical model of the TNF α -mediated NF- κ B-signaling pathway.

generation or raise the rate of degradation for the enzyme, the response would be slowed down. This implies that the steady state of an enzyme (or protein) can modulate the response of the signal.

TNF α -Mediated NF- κ B-Signaling Pathway

TNF α is a potent cytokine often produced by various cell types including macrophage, monocytes, lymphocytes, keratinocytes, and fibroblasts in response to inflammation, infection, injury, and other environmental challenges (Veronique and Michael 2001). Exposure of cells to TNF α can recruit an activation of a caspase cascade that leads to apoptosis. However, more commonly, the binding of TNF α to its receptor causes an activation of two major transcriptional factors, AP-1 and NF- κ B; these, in turn, induce genes involved in chronic and acute inflammatory responses. NF- κ B is regulated primarily by phosphorylation of inhibitory proteins, the I κ Bs, which are retained in the cytoplasm of nonstimulated cells. In response to TNF α and other agonists, the I κ Bs are phosphorylated by the I κ B kinase (IKK) complex, resulting in their ubiquitination, degradation, and the nuclear translocation of the freed NF- κ B. Generally, TNF α exerts its effects through two distinct receptors, TNFR1 and TNFR2. Binding of the inherently trimeric TNF α to TNFR1 induces receptor trimerization and recruitment of several signaling proteins to the cytoplasmic domains of the receptors as shown in Figure 6. The first protein recruited to TNFR1 is TNFR1-associated death-

domain protein (TRADD), which serves as a platform to recruit at least three additional mediators, a receptor-interacting protein (RIP), a Fas-associated death-domain protein (FADD), and a TNF-receptor-associated factor 2 (TRAF2). TRAF2 plays a central role in early events, common to TNFR1 and TNFR2, which leads to the IKK activation (Veronique and Michael 2001).

Figure 6 shows the graphical model of the TNF α -mediated NF- κ B-signaling pathway.

Mathematical Modeling of TNF α -Mediated NF- κ B-Signaling Pathway

Applying the previous complemented model 7 of the reaction kinetics to each pathway step-by-step and integrating each model into a whole, the final mathematical model of the TNF α -mediated NF- κ B-signaling pathway, on the basis of a set of ODEs, is summarized in the Appendix. Moreover, all of the relevant definitions of variables and parameters appearing in the ODE model, together with nominal values for simulation studies in the Simulation Studies section, are given in Tables 1 and 2. The terminology of the dependent variables and their acronyms follows the conventional biological description, for which the symbols are based on common use of mathematical notation. It is assumed that the cell keeps the concentration of each signaling protein constant before and after each signaling, that is, the concentration of these proteins returns to a steady state after the reaction. Note that this assumption reflects

the biological observation. In addition, we assume that the signaling pathway behaves as a slowly time-varying system as described in the Preliminaries section.

The exact value of parameters, such as the concentration of each signaling protein, the rate constants for the generation and degradation, etc., are difficult to obtain because their numerical value not only depends on the species and tissue, but also on the physiological state of the cells/organism. Hence, we derive the nominal value for each parameter from the reaction time. For

Table 2. Definition of Variables and Nominal Values for Parameters in Reaction Kinetics

Independent variables	Symbol	$[\mu\text{M}^{-1}\text{sec}^{-1}]/[\text{sec}^{-1}]$
TNF/TNFR1 ratio	k_{12_3}/k_{3_12}	0.00096/0.004
TNFR1/TRADD ratio	k_{34_5}/k_{5_34}	0.00096/0.004
TRADD/FADD ratio	k_{56_9}/k_{9_56}	0.00096/0.004
TRADD/RIP ratio	k_{57_10}/k_{10_57}	0.00096/0.004
RIP/TRAF2	k_{810_11}/k_{11_810}	0.00096/0.004
TRAF2/IKK ratio	k_{911_12}/k_{12_911}	0.00096/0.004
IKK activation ratio	k_{12_13}	0.1
IKK_act/(I κ B/NF- κ B) ratio	k_{1314_15}/k_{15_1314}	0.00096/0.004
NF- κ B activated ratio	k_{15_1617}	0.00096/0.004
Degradation ratio	k_{16}	0.1

Modeling the NF- κ B-Signaling Pathway

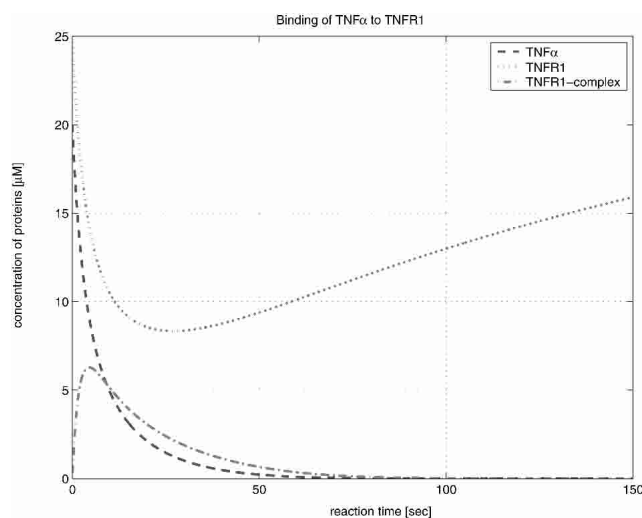


Figure 7 Binding of TNF α to TNFR1.

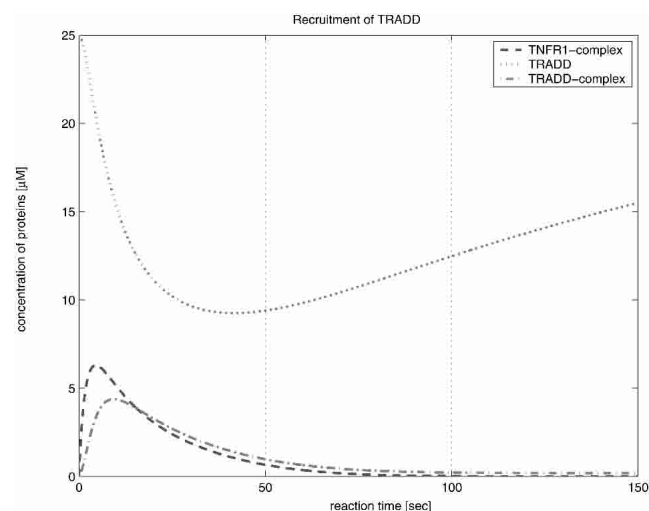


Figure 8 Recruitment of TRADD.

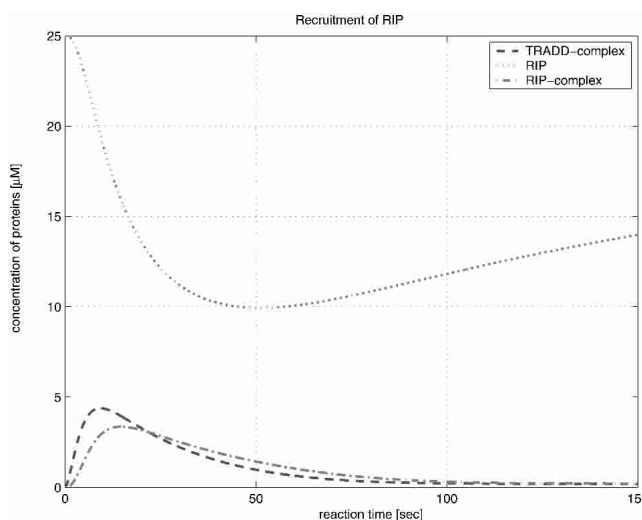


Figure 9 Recruitment of RIP.

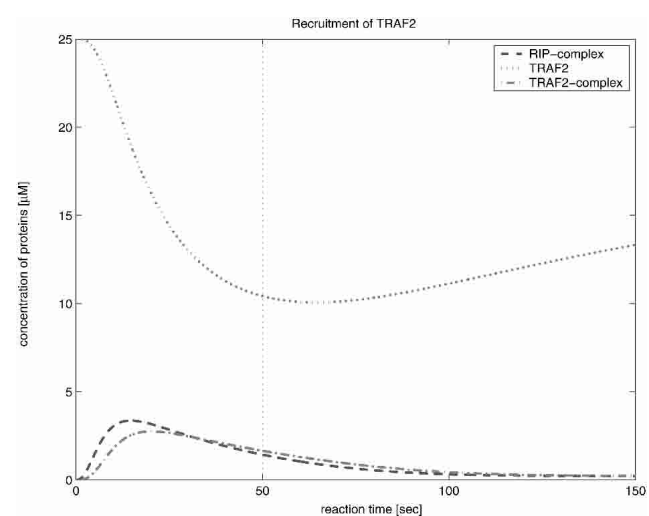


Figure 10 Recruitment of TRAF2.

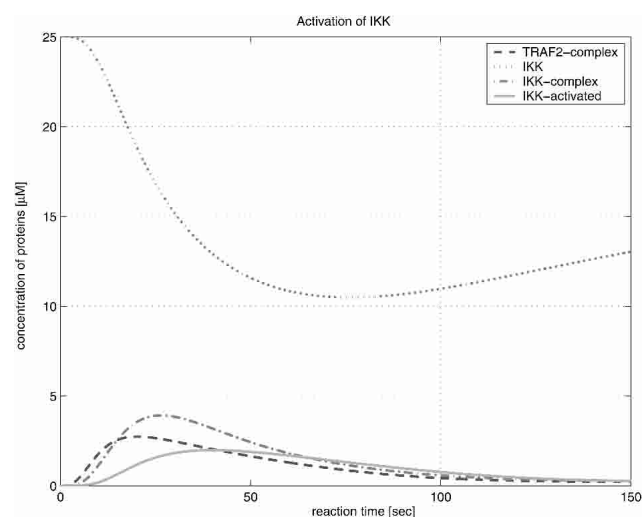


Figure 11 Activation of IKK.

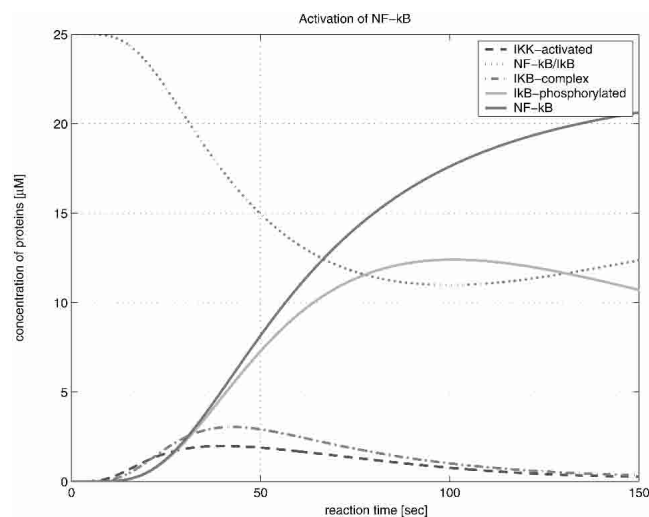


Figure 12 Activation of NF- κ B.

instance, it is well known that it takes several seconds for two proteins to interact in the signaling pathway. We therefore make a reasonable quantitative inference from the reaction time. In this way, we summarize the derived nominal value for each parameter in Tables 1 and 2. Note that whenever we can get more specific information regarding the reaction time in a particular experimental condition, for example, the measured time for TNF α returning to the steady state after a reaction from its initial state, we can then modify more exact parameter values for the rate of generation, the rate of degradation, and so forth, on the basis of new information about the reaction time.

In Table 2, the rate of formation for the complex is assumed to be $0.00096 \text{ } (\mu\text{M}^{-1}\text{sec}^{-1})$, the rate of dissociation from the complex is assumed to be $0.004 \text{ } (\text{sec}^{-1})$, and the rate of production is assumed to be $0.1 \text{ } (\text{sec}^{-1})$.

Simulation Studies

The computer simulation is carried out with a 2GHz Pentium 4 PC. The differential equations in the model of the signaling pathway are solved by utilizing MATLAB library functions (ode45, etc.). The results can be classified into two groups as follows: one group for protein interactions, and the other group for three-dimensional profiles along with time and the concentration for TNF α . The latter group is to be called the sensitivity to TNF α .

As in the Preliminaries and the Mathematical Modeling sections, it is assumed that the cell keeps the concentration of each signaling protein constant before and after signaling takes place. Figure 7–12 illustrates the formation process for the complex from TNF α and TNFR1. The concentration of the complex reaches its maximum in about 10 sec, and then decreases exponentially to the initial condition zero. TNF α reduces to $\sim 8 \text{ } \mu\text{M}$, and then increases gradually. After a while, it returns to its steady state. Here, we consider the change of concentration of the complex as a signal. This signal is transferred to the next stage in Figure 8, in which TRADD is recruited and bound to the TNFR1 complex, forming a TRADD complex. The concentration change of the TRADD complex can be considered as another biochemical signal transferred from the previous stage. Similarly, the signal is transferred to IKK in Figure 11, below (for intermediate procedures, refer to Figs. 8–11). The signal activates IKK, and then the activated IKK finally phosphorylates I κ B, which, in turn, releases NF- κ B. Thereupon, the concentration of released NF- κ B increases to its maximum value as illustrated in Figure 12. Finally, the released NF- κ B translocates into the nucleus and initiates transcription of the gene that responds to the variation of the ligand TNF α .

The second group of the results is the three-dimensional graphical representation that illustrates the profile as function time and the concentration of TNF α in Figure 13. The concentration varies from 0 to $35 \text{ } \mu\text{M}$. From the three-dimensional representation, we obtain information about the dynamics of signaling components involved in the pathway, dependent on variations of the initial TNF α concentration. In Figure 14, TNFR1 is very quickly ($<20 \text{ sec}$) occupied by TNF α at its highest density, and then, after a while, TNFR1 turns to its steady state; however, it takes a longer period of time to recover its steady state. This implies that TNFR1 becomes more desensitized to the external stimulation as the initial TNF α concentration increases. That is, if the same stimulation is given consecutively to the cell receptor, then a lower signal transduces through the pathway. Figure 15 shows the profile of TNFR1-complex, in which the concentration increases along with time and the concentration of TNF α . This shows that TNFR1 also becomes desensitized to the external stimulation, as the number of receptor molecules trapped in TNFR1 complex becomes larger as the initial TNF α concentration

is higher. The TNF α complex recruits TRADD in Figure 16, and its concentration decreases during the recruiting, whereas the concentration of the TRADD complex increases along with time and the concentration of TNF α as shown in Figure 17. The increase of the TRADD complex, anchored in the upstream-signaling component, implies that the apoptosis mechanism and the cell-proliferation mechanism become activated in accordance with the increase of the initial TNF α concentration, as TRADD serves as a platform in which FADD (responsible for apoptosis) and RIP (responsible for proliferation) are recruited as shown in Figure 18. However, it depends on the internal mechanism as to whether the apoptosis or the proliferation mainly affects the fate of the cell. Figures 19–23 illustrate the dynamics of the activated IKK. We note that the concentration of IKK increases only about 2.2 times, whereas the initial TNF α concentration increases 35 times. This indicates that IKK is relatively insensitive to the change of the initial TNF α concentration. Figures 24–27 show the signaling profile of the free NF- κ B in a cytoplasm released from the NF- κ B/I κ B complex. We can also find the insensitive change of the free NF- κ B in a cytoplasm according to the variation of the initial TNF α concentration. That is, we note that the NF- κ B concentration varies ~ 2.5 times as the initial TNF α concentration increases 35 times. The sensitivity characteristics of the signaling components with respect to the initial TNF α concentration (i.e., input to the system) is related to the stability and robustness of the signal transduction system in view of systems theory.

Another interesting observation from these three-dimensional simulation results relates to the signal propagation delay. We note that the time delay of each component in the pathway increases for downstream components as shown in the contours depicted in the bottom planes of Figures 14, 16, 18, 20, and 24. The signal propagation delay of the pathway with multiple inputs (i.e., different kinds of stimulations and corresponding receptors, etc.) can result in various dynamics such as signaling discontinuity through merging and amplification. Figs. 13–27

DISCUSSION

In this work, we have investigated a system-theoretic approach to analysis and quantitative modeling of the TNF α -mediated NF- κ B-signaling pathway. The quantitative model was complemented with a qualitative model. We proposed an intuitive graphical representation of the signaling pathway on the basis of a bipartite-directed multigraph, and then presented the mathematical model of the signal pathway via a set of ODEs on the basis of complemented reaction kinetics. For simulation studies, we have derived the nominal value for each parameter through inference on the basis of reaction time. The simulation studies have illustrated the process of variation of each protein concentration along with the TNF α -mediated NF- κ B-signaling pathway as the concentration of the ligand TNF α varies. Moreover, the computer simulation, on the basis of the proposed quantitative model, has revealed the transient behavior of the signaling pathway. The proposed signaling-pathway model based on a system-theoretic approach can be extended and is applicable to other signaling pathways in the same manner. As a next step, it will be useful to include more detailed elements of the pathway excluded in the current study. This may lead to hybrid systems, combining discrete event systems with continuous dynamics in order to model the switching or decision making in signaling pathways. Feedback control and time delays are further challenges for an extended model. As yet, the model produces hypotheses and can guide the biologist in experimental design. To predict quantitative behavior from the proposed model and to apply the predicted results in biotechnological application, the proposed

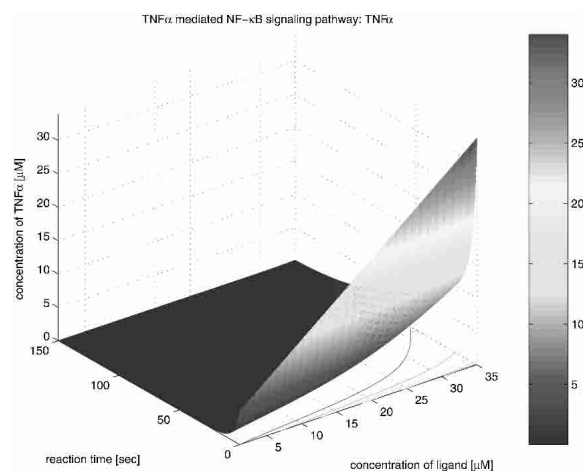


Figure 13 TNF α .

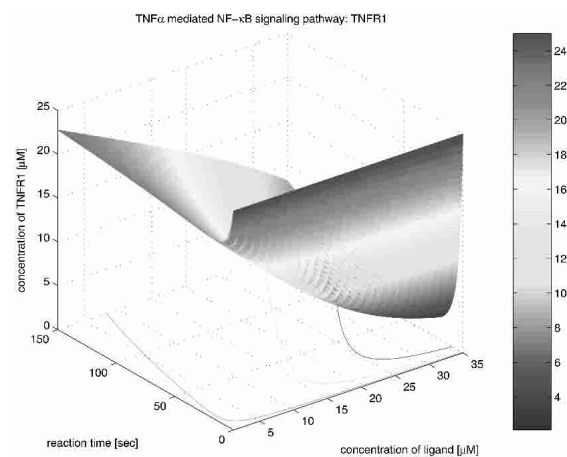


Figure 14 TNFR1.

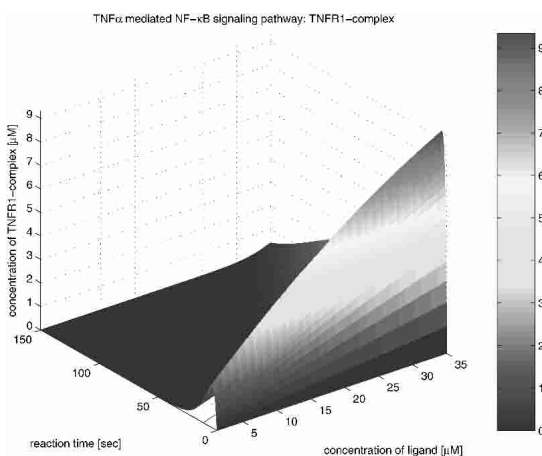


Figure 15 TNFR1 complex.

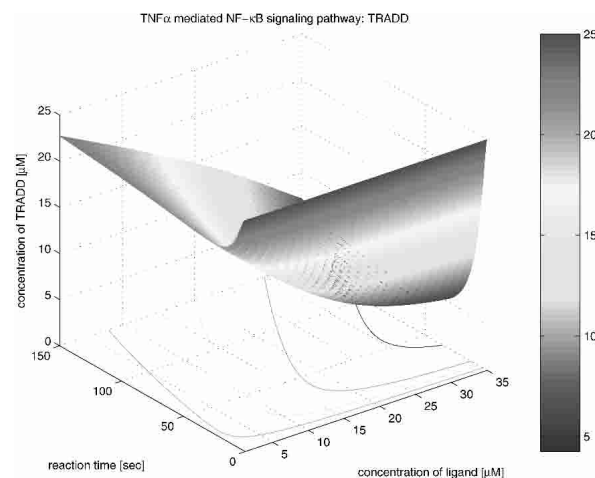


Figure 16 TRADD.

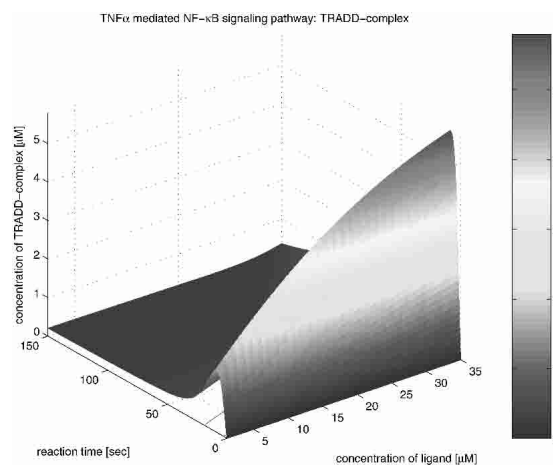


Figure 17 TRADD complex.

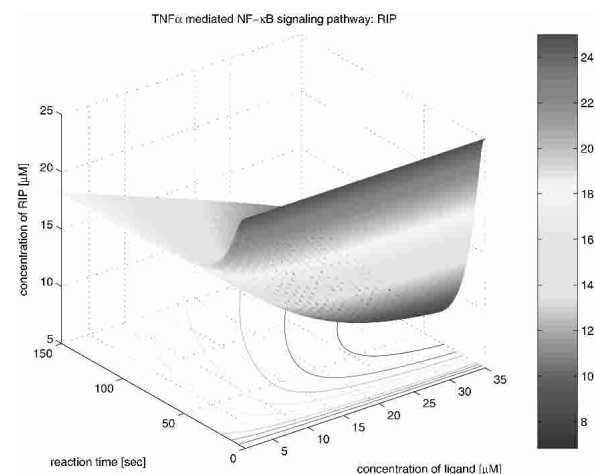


Figure 18 RIP.

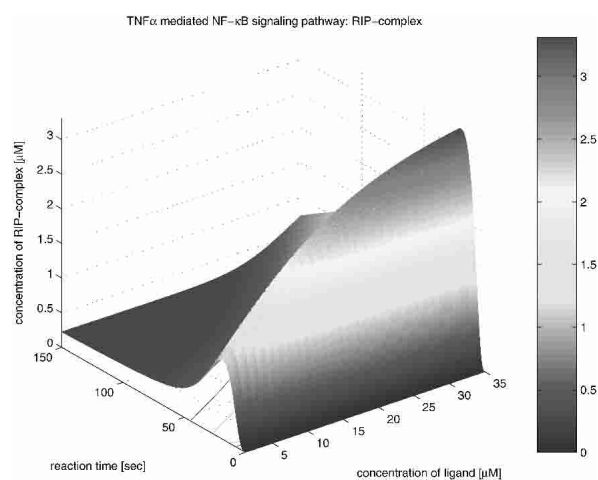


Figure 19 RIP complex.

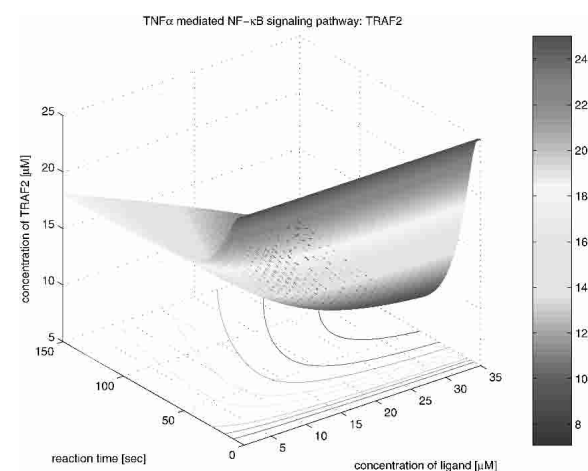


Figure 20 TRAF2.

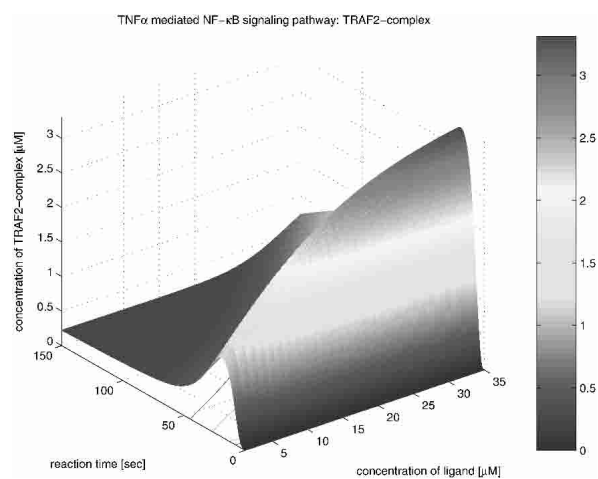


Figure 21 TRAF2 complex.

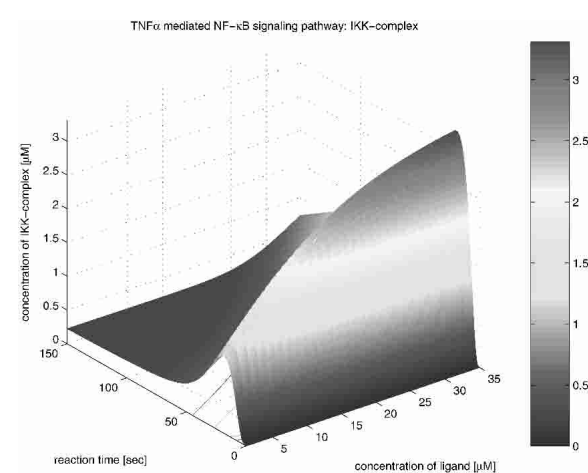


Figure 22 IKK complex.

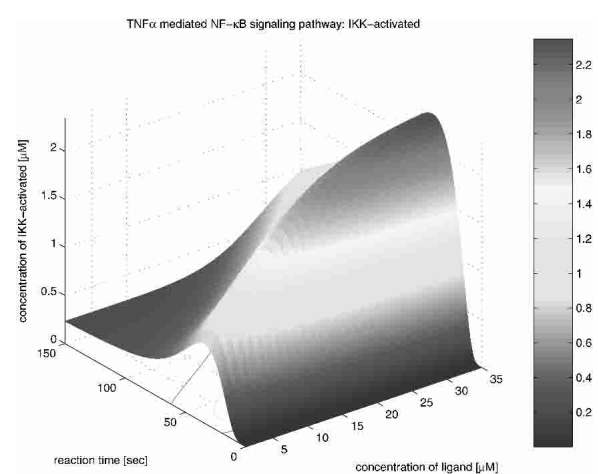


Figure 23 Activated IKK.

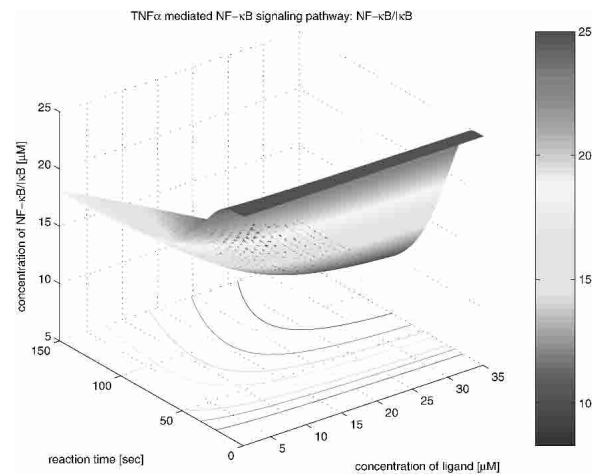
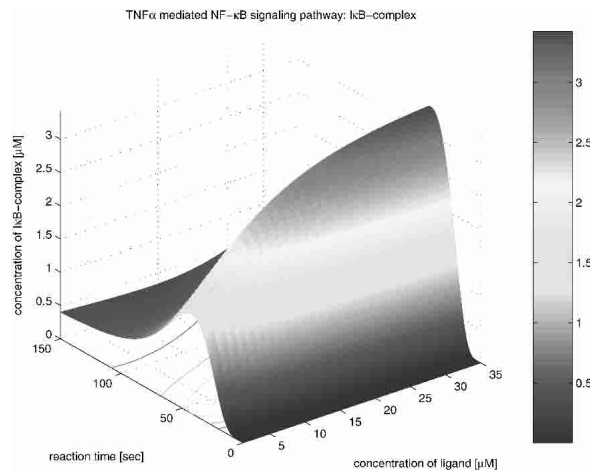
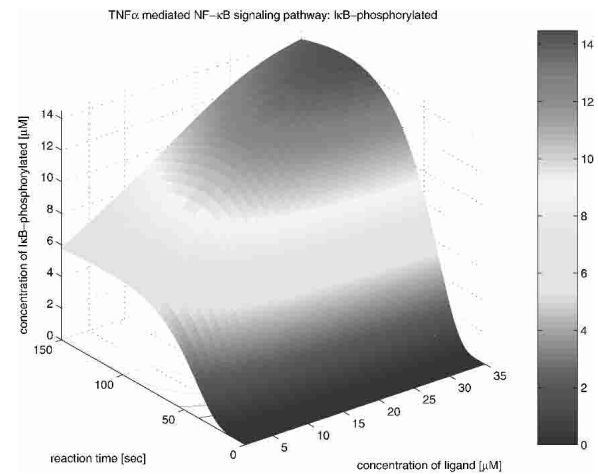
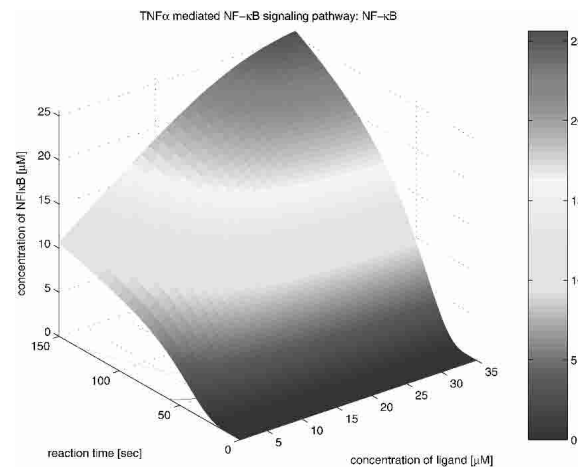


Figure 24 NF κ B/I κ B.

**Figure 25** I κ B complex.**Figure 26** Phosphorylated I κ B.**Figure 27** NF κ B.

model needs to be supplemented by more accurate nominal values for each parameter on the basis of *in vivo* experiments. The experimental validation of the quantitative predictions made by the simulation studies also remains as important further research.

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APPENDIX

The mathematical model of the TNF α -mediated NF- κ B-signaling pathway on the basis of complemented reaction kinetics. This appendix summarizes the mathematical model of the TNF α -mediated NF- κ B-signaling pathway on the basis of the comple-

mented reaction kinetics in the Preliminaries section. For the nominal value of each parameter, refer to Tables 1 and 2.

$$\frac{dm_1(t)}{dt} = k_{12,3} \cdot m_1(t) \cdot m_2(t) + k_{3,12} \cdot m_3(t)$$

$$\frac{dm_2(t)}{dt} = k_{2f} - k_{2r} \cdot m_2(t) - k_{12,3} \cdot m_1(t) \cdot m_2(t) + k_{3,12} \cdot m_3(t)$$

$$\begin{aligned} \frac{dm_3(t)}{dt} = & k_{12,3} \cdot m_1(t) \cdot m_2(t) - [k_{3,12} + k_{34,5} \cdot m_4(t)] \cdot m_3(t) \\ & + k_{5,34} \cdot m_5(t) \end{aligned}$$

$$\frac{dm_4(t)}{dt} = k_{5,34} \cdot m_5(t) - [k_{4f} + k_{34,5} \cdot m_3(t)] \cdot m_4(t) + k_{4f}$$

$$\begin{aligned} \frac{dm_5(t)}{dt} = & k_{34,5} \cdot m_3(t) \cdot m_4(t) - [k_{56,9} \cdot m_6(t) + k_{57,10} \cdot m_7(t) \\ & + k_{6f} + k_{5,34}] \cdot m_5(t) + k_{9,56} \cdot m_{18}(t) + k_{10,57} \cdot m_{10}(t) \end{aligned}$$

$$\frac{dm_6(t)}{dt} = [k_{6f} + k_{56,9} \cdot m_5(t)] \cdot m_6(t) + k_{9,56} \cdot m_{18}(t)$$

$$\frac{dm_7(t)}{dt} = k_{7f} - [k_{7r} + k_{57_{10}} \cdot m_5(t)] \cdot m_7(t) + k_{10_{57}} \cdot m_{10}(t)$$

$$\frac{dm_8(t)}{dt} = [k_{8r} + k_{810_{11}} \cdot m_{10}(t)] \cdot m_8(t) + k_{11_{810}} \cdot m_{11}(t) + k_{8f}$$

$$\frac{dm_9(t)}{dt} = [k_{9r} + k_{911_{12}} \cdot m_{11}(t)] \cdot m_9(t) + k_{12_{911}} \cdot m_{12}(t) + k_{9f}$$

$$\frac{dm_{10}(t)}{dt} = k_{57_{10}} \cdot m_5(t) \cdot m_7(t) - k_{810_{11}} m_8(t) \cdot m_{10}(t) + k_{11_{810}} \cdot m_{11}(t)$$

$$\frac{dm_{11}(t)}{dt} = k_{810_{11}} \cdot m_8(t) \cdot m_{10}(t) + k_{12_{911}} \cdot m_{12}(t) - k_{911_{12}} \cdot m_9(t)$$

$$\frac{dm_{12}(t)}{dt} = k_{911_{12}} \cdot m_9(t) \cdot m_{11}(t) - [k_{12_{13}} + k_{12_{911}}] \cdot m_{12}(t) + k_{11_{810}} \cdot m_{11}(t)$$

$$\frac{dm_{13}(t)}{dt} = k_{15_{1314}} \cdot m_{15}(t) + k_{12_{13}} \cdot m_{12}(t) - k_{1314_{15}} \cdot m_{13}(t) \cdot m_{14}(t)$$

$$\frac{dm_{14}(t)}{dt} = k_{14f} + k_{15_{1314}} \cdot m_{15}(t) - [k_{14r} + k_{1314_{15}} \cdot m_{13}(t)] \cdot m_{14}(t)$$

$$\frac{dm_{15}(t)}{dt} = k_{1314_{15}} \cdot m_{13}(t) \cdot m_{14}(t) - [k_{15_{1314}} + k_{15_{1617}}] \cdot m_{15}(t)$$

$$\frac{dm_{16}(t)}{dt} = k_{15_{1617}} \cdot m_{15}(t) - k_{16} \cdot m_{16}(t)$$

$$\frac{dm_{17}(t)}{dt} = k_{15_{1617}} \cdot m_{15}(t)$$

$$\frac{dm_{18}(t)}{dt} = k_{56_{9}} \cdot m_5(t) \cdot m_6(t) - k_{9_{56}} \cdot m_{18}(t)$$

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