

Abstract

We propose a model based on the logistic equation and linear kinetics to study the effect of toxicants with various initial concentrations on a cells' population. To efficiently estimate the model's parameters, we design an Expectation Maximization algorithm. The model is validated by showing that it accurately represents the information provided by in-vitro experiments.

Introduction

At the Alberta Centre of Toxicology the effect of various toxicants on growth/death and morphology of human cells is investigated using the xCELLigence Real-Time Cell Analysis High Throughput in vitro assay. The cell index is measured as a proxy for the number of cells, and for each test substance in each cell line, time-dependent concentration response curves (TCRCs) are generated. The toxicants are grouped in clusters, according to the mode of action. The goal of this study is to find a model that could accurately reproduce these curves.

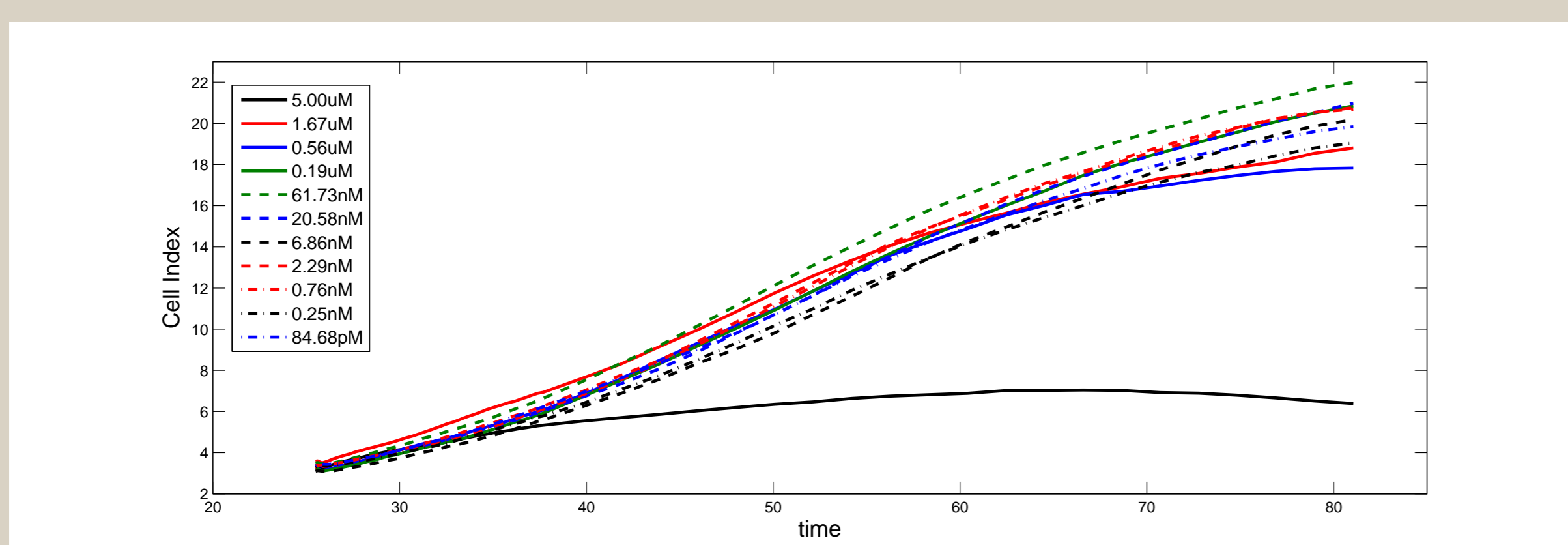


Figure 1: TCRCs for the toxicant PF431396

The Model

$$\begin{aligned} \frac{dn(t)}{dt} &= \beta n(t) \left(1 - \frac{n(t)}{K}\right) - \alpha C_0(t) n(t), & n(t): \text{cell index at time } t \\ \frac{dC_0(t)}{dt} &= \lambda_1^2 CE(t) - \eta_1^2 C_0(t), & C_0(t): \text{internal toxicant concentration} \\ \frac{dCE(t)}{dt} &= \lambda_2^2 C_0(t) n(t) - \eta_2^2 CE(t) n(t), & CE(t): \text{external toxicant concentration} \end{aligned}$$

β : cell growth rate
 K : capacity volume
 λ_1^2, η_1^2 : toxicant uptake and input rates from environment
 λ_2^2, η_2^2 : toxicant uptake and loss rates from cells

To estimate the parameters we write the system in a state-space form. From the experimental data recorded in the TCRCs we get observations, possibly affected by measurement errors, only for the cell index $n(t)$. Using the Euler integration scheme with time step h , we get the discrete state-space system:

$$\begin{aligned} x_{k+1} &= x_k + h \begin{bmatrix} \beta x_k[1] \left(1 - \frac{x_k[1]}{K}\right) - \alpha x_k[2] x_k[1] \\ \lambda_1^2 x_k[3] - \eta_1^2 x_k[2] \\ \lambda_2^2 x_k[2] x_k[1] - \eta_2^2 x_k[3] x_k[1] \end{bmatrix} + v_{k+1} \\ y_{k+1} &= C x_{k+1} + w_{k+1} \end{aligned}$$

- ▶ $x_{k+1} = [n, C_0, CE]^t$ is the state of the system
- ▶ y_{k+1} is the observation at time step $k + 1$
- ▶ v_k and w_k , are uncorrelated, $v_k \sim N(0, Q)$ and $w_k \sim N(0, R)$
- ▶ h is the time step; $C = [1 \ 0 \ 0]$; $x_1[1] = n(0)$, $x_1[2] = C_0(0) = 0$ and $x_1[3] = CE(0)$ comes from the measured data at $t_0 = 0$

Parameter Estimation

- ▶ Using the experimental data corresponding to the negative control (no toxicant) we can estimate β and K using the nonlinear least square method based on the analytic solution of the logistic equation.
- ▶ The remaining parameters $\Theta = \{Q, R, \alpha, \lambda_1, \lambda_2, \eta_1, \eta_2\}$ are estimated using the state-space model and the Expectation Maximization (EM) algorithm based on the unscented filter (UF).
- ▶ The UF is an alternative to the extended Kalman filter (EKF) to calculate the filtered values $\bar{x}_i = E[x_i | y_1, \dots, y_i]$, $\bar{P}_i = E[(x_i - \bar{x}_i)(x_i - \bar{x}_i)^t | y_1, \dots, y_i]$, the predicted values $\hat{x}_{i+1} = E[x_{i+1} | y_1, y_2, \dots, y_i]$, $\hat{P}_{i+1} = E[(x_{i+1} - \hat{x}_{i+1})(x_{i+1} - \hat{x}_{i+1})^t | y_1, \dots, y_i]$ and the smoothed values $x_{i|N} = E[x_i | y_1 \dots y_N]$ and $P_{i|N} = E[(x_i - x_{i|N})(x_i - x_{i|N})^t | y_1 \dots y_N]$ based on the available observations $y_1 \dots y_N$.
- ▶ The state equation is non-linear, so an approximation is needed during the E-step. The likelihood and the conditional likelihood $P(x_1, \dots, x_N, y_1, \dots, y_N | y_1, \dots, y_N)$ are approximated based on a linearization of the state equation.

The EM Algorithm

Initialize the model parameters $\Theta = \{Q, R, \alpha, \lambda_1, \lambda_2, \eta_1, \eta_2\}$

Repeat until the log likelihood has converged

The E step: compute $\hat{E} = E[\log P(x_1, \dots, x_N, y_1, \dots, y_N) | y_1, \dots, y_N]$
 For $k=1$ to N

Run the UF filter to compute \bar{x}_{k+1} , \bar{P}_{k+1} , \hat{x}_{k+1} , \hat{P}_{k+1} and $\bar{P}_{x_k x_{k+1}}$

For $k=N$ to 1

Calculate the smoothed values $x_{k|N}$, and $P_{k|N}$

The M step

Update the values of the parameters Θ to maximize \hat{E}

Model Validation

- ▶ We divide the experimental data into a training set (around 70% of the data), and a test set. We use the N observations in the training set to estimate the parameters.
- ▶ Once the parameters are estimated, we can predict the future values of x_i , $i = N + 1, N + 2, \dots$. We validate the model by comparing these predictions with the experimental data in the test set.

Table 1: Estimated Values of Parameters

Toxicant	Cluster	β	K	η_1	λ_1	λ_2	η_2	α
monastrol	X	0.074	18.17	0.209	0.177	0.204	0.5	0.016
HA1100 hydrochloride	I	0.077	21.913	0.143	0.0098	0.0786	0.147	0.351

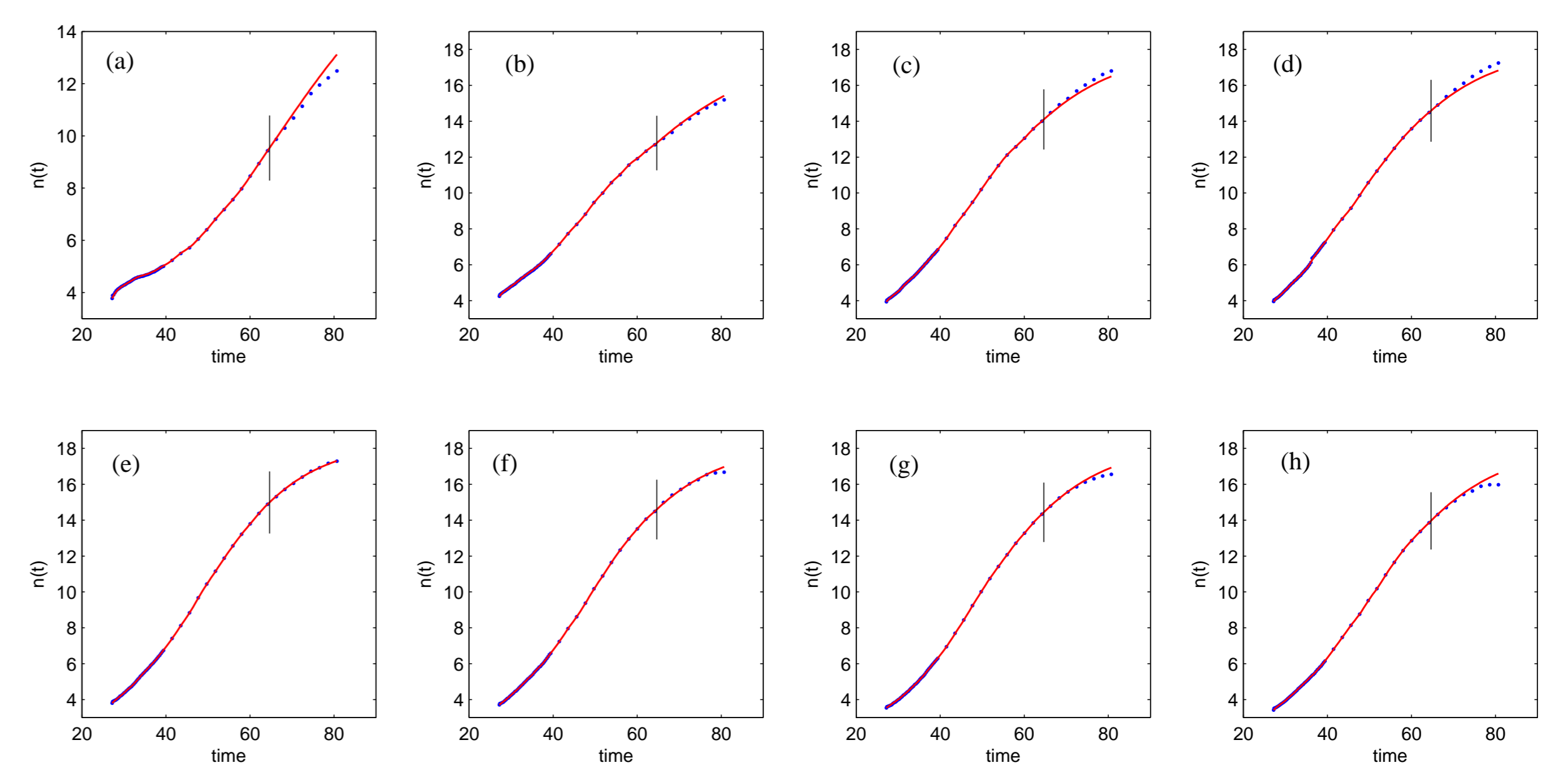


Figure 2: Estimation results for monastrol: dot for experimental data, line for filtered or predicted observations; (a) $CE(0)=100.00\mu\text{M}$, (b) $CE(0)=33.33\mu\text{M}$, (c) $CE(0)=11.11\mu\text{M}$, (d) $CE(0)=3.70\mu\text{M}$, (e) $CE(0)=1.23\mu\text{M}$, (f) $CE(0)=0.41\mu\text{M}$, (g) $CE(0)=0.14\mu\text{M}$, (h) $CE(0)=45.72\text{nM}$

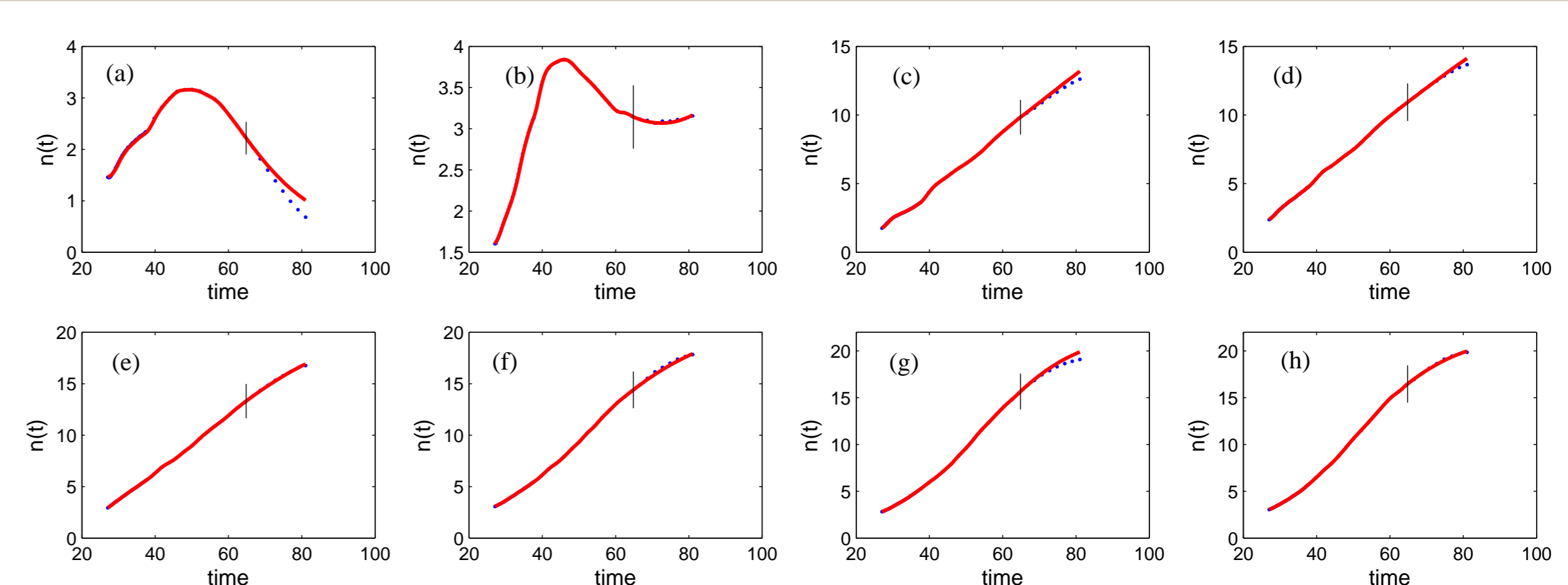


Figure 3: Estimation results for HA1100 hydrochloride: dot for experimental data, line for filtered or predicted observations; (a) $CE(0)=1.00\text{mM}$, (b) $CE(0)=0.33\text{mM}$, (c) $CE(0)=0.11\text{mM}$, (d) $CE(0)=37.04\mu\text{M}$, (e) $CE(0)=12.35\mu\text{M}$, (f) $CE(0)=4.12\mu\text{M}$, (g) $CE(0)=1.37\mu\text{M}$, (h) $CE(0)=0.46\mu\text{M}$

Conclusions

- ▶ The proposed mathematical model is in good agreement with the experimental TCRCs.
- ▶ The EM algorithm based on the unscented filter gives accurate predictions of the concentration of toxicant outside the cells.
- ▶ The model can be used to determine an appropriate range for the initial concentration of chemicals $CE(0)$ used in the experiments such that both values smaller and larger than the threshold between extinction and persistence are included.

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