Models in Toxicology

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ALL MODELS ARE WRONG AND SOME ARE USE(FUL)/(LESS)

Mathematical Model

To describe real world processes using a simple language, i.e. mathematical expression



The two most important mechanisms in **T**oxicology

- DOSE-RESPONSE RELATIONSHIP a correlation between dose of a toxic substance administered or received and the incidence of an adverse (including health) effect in exposed population.
- **TOXICOKINETICS** process of uptake of toxicants by the body of organism, the biotransformation they undergo, the distribution of the toxicants and their metabolites in the tissues, and the elimination of the toxicants and their metabolites from the body.



Dose-Response Models

Dose-response relationship (DRR) is fundamental to toxicology. Understanding the association between effect and dose is the basis of safety evaluation, assuming that the effect is the result of the substance administered. In toxicology, we can observe a <u>quantal</u> response (mortality or number of animals affected) and a graded or continuous response (weight, enzyme activity, etc).

When a sufficient number doses is used in an experiment with a sufficient number of animals, the result can be represented by a sigmoid dose-response curve. In classical analysis of dose-response data, probit or logit transformation is used to transform the sigmoid curve into a linear curve. Based on the slope and intercept of this linear curve, the values of LD₅₀ or ED₅₀ can then be derived.



Alternatively, a logistic response model can be used to describe the dose-effect relationship:

$$Y = \frac{c}{1 + e^{b(X-a)}} \tag{18}$$

where Y is the observed biological rate, X is the (natural) logarithm of the concentration, c is the undisturbed level of the biological rate, a is the logarithm of the concentration at which the biological rate is half of the undisturbed level, and b is a slope parameter.

Parameter **b** indicates the rate of increase of inhibition with increasing concentration around ED₅₀ concentration. The higher the value of **b**, the more abruptly the biological rate decreases, which would mean that the toxicant strongly acts on the organism. On the other hand a low value of b does not necessarily mean that a toxicant does not act strongly on the organism.



TOXICOKINETICS MODEL

- Compartment-based models describe toxicant movement between compartments.
- A compartment represents the amount of a compound that behaves as though it exists in a homogeneously well-mixed container and moves across the compartment boundary with a single uptake or elimination rate coefficient.





Compartment Models

- A simple compartment model containing water and organism compartments. **The water represents the source**
- Of toxicant and the organism represents the toxicant sink.



ASUMPTIONS:

- The toxicant is well mixed and homogeneous within each compartment
- No compound biotransformation occurs
- The uptake rate constants and clearances remain constant over time (if the organism undergoes physiological change, this assumption can be violated)
- The transfer between compartment is first order. Thus, the flux across the boundary depends on the chemical activity (concentration) in the respective compartment. The net flux is the sum of the uptake and loss fluxes across the compartment boundaries



The model relates the amount or concentration of a compound in one compartment with that in another:



(1)

Where

- C_a = the concentration of the chemical in the organism (mol/kg)
- C_w = the concentration of the chemical in the water (mol/L)
- k_u = the uptake rate constant (L/kg.d)
- k_e = the elimination rate constant (1/d) t = time (d)



If **Cw** is held constant, as ideally occurs in flow-through experiments and is often assumed for field exposures, Equation (1) can be exactly integrated to yield

$$Ca = \frac{k_u \cdot Cw}{k_e} (1 - e^{-k_e t})$$

The uptake rate constant can be derived from the initial uptake of the chemical by the organism, when elimination is asumed to be negligible

$$C_a = k_u C_w t$$

$$\frac{dCa}{dt} = (k_u \cdot Cw) - (k_e \cdot Ca)$$
⁽¹⁾

Where

- C_a = the concentration of the chemical in the organism (mol/kg)
- C_w = the concentration of the chemical in the water (mol/L)
- k_u = the uptake rate constant (L/kg.d) k_e = the elimination rate constant (1/d) t = time (d)

$$Ca = \frac{k_u \cdot Cw}{k_e} (1 - e^{-k_e t})$$



 $dC_a/dt = 0$

--- steady state

$$BCF = C_a/C_w = K_u/K_e$$

Biological Concentration Factor



Typical Experimental Data

t	Ca
0	2.00
1	4.50
2	6.70
3	8.90
4	12.10
5	15.60
6	22.70
7	28.90
8	29.10
9	29.90
10	29.90
11	29.80
12	29.90
13	29.85
14	29.95
15	29.80
16	29.80
17	29.89





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MODEL DEFINITION $Ca = \frac{k_u \cdot Cw}{k_e} (1 - e^{-k_e t})$



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ESTIMATION PROCEDURE (algorithms, iterations, sensitivity)

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Iteration process

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Ku(o) = ku optimum



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SAVING PREDICTED VALUES & RESIDUALS

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	2 134.1074970 3.01014910 .171090240
	3 143.7823735 5.54496196 .160166809
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	4 143.6848572 5.61031957 .163439697
	4.1 143.6805519 5.59395216 .162688821
	5 143.6805519 5.59395216 .162688821
	5.1 143.6803067 5.59786037 .162867354
	6 143.6803067 5.59786037 .162867354
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	\Rightarrow Run stopped after 14 model evaluations and 7 derivative evaluations.
	Iterations have been stopped because the relative reduction between success

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	Regression 2 10619.33681 5309.66840					
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Parameter Estimation (SPSS for Windows 10.0)

• Il the derivatives will be calculated numerically.

•	Iteration	Residual SS	KU	KE
•	1	144.2117274	5.50000000	.160000000
•	1.1	143.6859841	5.61110738	.163502397
•	2	143.6859841	5.61110738	.163502397
•	2.1	143.6806092	5.59362163	.162673678
•	3	143.6806092	5.59362163	.162673678
•	3.1	143.6803099	5.59793821	.162870920
•	4	143.6803099	5.59793821	.162870920
•	4.1	143.6802932	5.59692098	.162824358
•	5	143.6802932	5.59692098	.162824358
•	5.1	143.6802923	5.59716166	.162835369

- Run stopped after 10 model evaluations and 5 derivative evaluations.
- Iterations have been stopped because the relative reduction between successive residual sums of squares is at most SSCON = 1.000E-08



Curve Fitting



Figure 1. Dynamics of chemical concentration in organism

Iteration	Residual SS	KU	KE
1	6606.044124	6.00000000	2.00000000
1.1	2.3679+301	-54.742782	-24.677153
1.2	89301381745	12.7747585	49618856
1.3	5693.251719	6.69601641	1.75940276
2	5693.251719	6.69601641	1.75940276
2.1	3989.165430	7.91880399	1.42238222
3	3989.165430	7.91880399	1.42238222
3.1	1579.618177	9.87011396	1.03883716
4	1579.618177	9.87011396	1.03883716
4.1	929.5788109	8.76624160	.591966797

Run stopped after 10 model evaluations and 4 derivative evaluations. The iterations limit has been reached.



Single compartment model with time-varying input

In the real world situation, constant exposure to toxicants is a very special case. Mostly, levels of toxic substances released into environment are highly variable. Constant exposure will be realized only in the case of persistent toxicants in a well-buffered environment, *e.g.* heavy metals in soil. In almost all other cases, exposure is not constant. This may vary from erratic fluctuations to peaks followed by a gradual decrease. Concentrations of environmental pollutants can be variable due to varying rates of input and dilution, changes in chemical form and solubility, and degradation.

For non-persistent chemicals, such as pesticides, the half-life, or degradation time, is a very important variable determining ecological effects. Exposure concentrations in toxicity tests are characterized by an initial peak at time zero, followed by a gradual decrease. There is no theoretical framework for dealing with these non-constant exposures in the standard statistical analysis of concentration-response experiments.



Various terms have been used to describe the patterns of time-variable exposures, including pulse, plug, spike, episodic, fluctuating and intermittent exposures. In general these patterns can be simplified into two types of variable exposure: (1) pulse exposure which involves one or more isolated and brief exposure periods, and (2) fluctuating exposure which can be defined as a continuous exposure to varying toxicant concentrations.

The present model deals with a pulse exposure followed by exponential decay ("diluted pulse"). This type of exposure is not an uncommon phenomenon, and can be found both in terrestrial or aquatic environments. In the case of metal contamination in aquatic environments, such as urban streams, a decreasing exposure can occur when chemical discharges were released intermittently during production processes, so there will be a dilution driven by the flow and volume of water in the streams.

$$C_{w}(t) = C_{w_0} e^{-k_0 t}$$

(13)



$$C_{w}(t) = C_{w_0} e^{-k_0 t}$$

where:

 Cw_0 = initial external concentration (e.g. in $\mu g/g$),

 $k_0 = rate constant$ for degradation of the chemical in the medium (e.g., in day-1).

The second assumption is that the kinetics of the concentration in the body follow a onecompartment model. This can be written as:

(13)

$$\frac{dC_a}{dt} = k_1 \underbrace{C_w}_{a}(t) - k_2 C_a(t) \tag{14}$$

where $C_a(t) =$ internal concentration at time t, $k_1 =$ rate constant for uptake, $k_2 =$ rate constant for elimination, and other loss processes from the body, such as metabolism.



$$\frac{dC_a}{dt} = k_1 \underbrace{C_w}_{t}(t) - k_2 C_a(t)$$
(14)

where $C_a(t)$ = internal concentration at time t, k_1 = rate constant for uptake, k_2 = rate constant for elimination, and other loss processes from the body, such as metabolism.

Most toxicity experiments start with animals transferred from a clean environment, so equation (2) can be integrated with the initial condition, $C_a(0) = 0$. Application of standard techniques (*e.g.* Laplace transforms, see Jacques, 1972), yields:

$$C_{a}(t) = \frac{k_{1}Cw_{0}}{k_{2} - k_{0}} \left(e^{-k_{0}t} - e^{-k_{2}t} \right)$$
(15)

(for the application, see Widianarko & van Straalen, 1996)



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