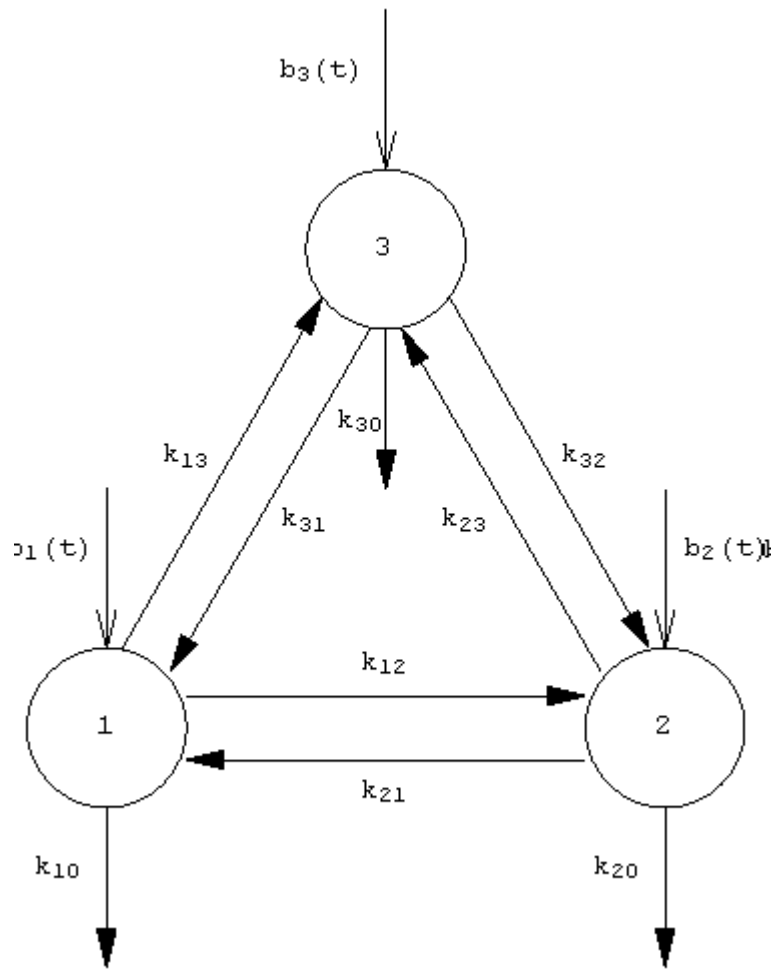


# Biokinectic modelling and Optimal design

British-Spanish Workshop on  
Optimal Design

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Compartmental analysis has applications in clinical medicine, pharmacokinetics, internal dosimetry, nuclear medicine, ecosystem studies and chemical reaction kinetics. It can be described as the analysis of a system in terms of compartments which separate the system into a finite number of component parts which are called compartments. Compartments interact through the exchange of species. Species may be a chemical substance, hormone, individuals in a population and so on.

$$\begin{pmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{pmatrix} = \begin{pmatrix} -(k_{10} + k_{12} + k_{13}) & k_{21} & k_{31} \\ k_{12} & -(k_{20} + k_{21} + k_{23}) & k_{32} \\ k_{13} & k_{23} & -(k_{30} + k_{31} + k_{32}) \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix} + \begin{pmatrix} b_1(t) \\ b_2(t) \\ b_3(t) \end{pmatrix}$$

# Compartmental general equation

$$\mathbf{x}'(t) = \mathbf{A} \mathbf{x}(t) + \mathbf{b}(t)$$

$$\mathbf{x}(t) = \mathbf{x}_0 e^{t\mathbf{A}} + \int_0^t \mathbf{b}(\tau) e^{(t-\tau)\mathbf{A}} d\tau$$

$$\mathbf{x}'(t) = [x'_1(t), x'_2(t), \dots, x'_n(t)]^T$$

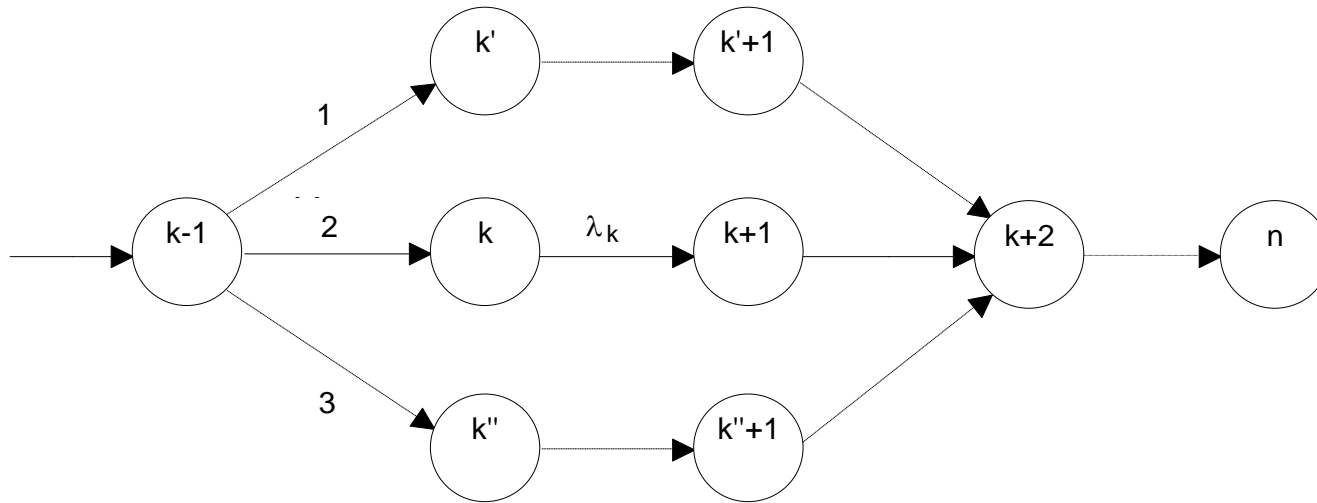
$$\mathbf{x}(t) = [x_1(t), x_2(t), \dots, x_n(t)]^T$$

$$\mathbf{b}(t) = [b_1(t), b_2(t), \dots, b_n(t)]^T$$

$$\mathbf{x}_0 = [x_1(0), x_2(0), \dots, x_n(0)]^T$$

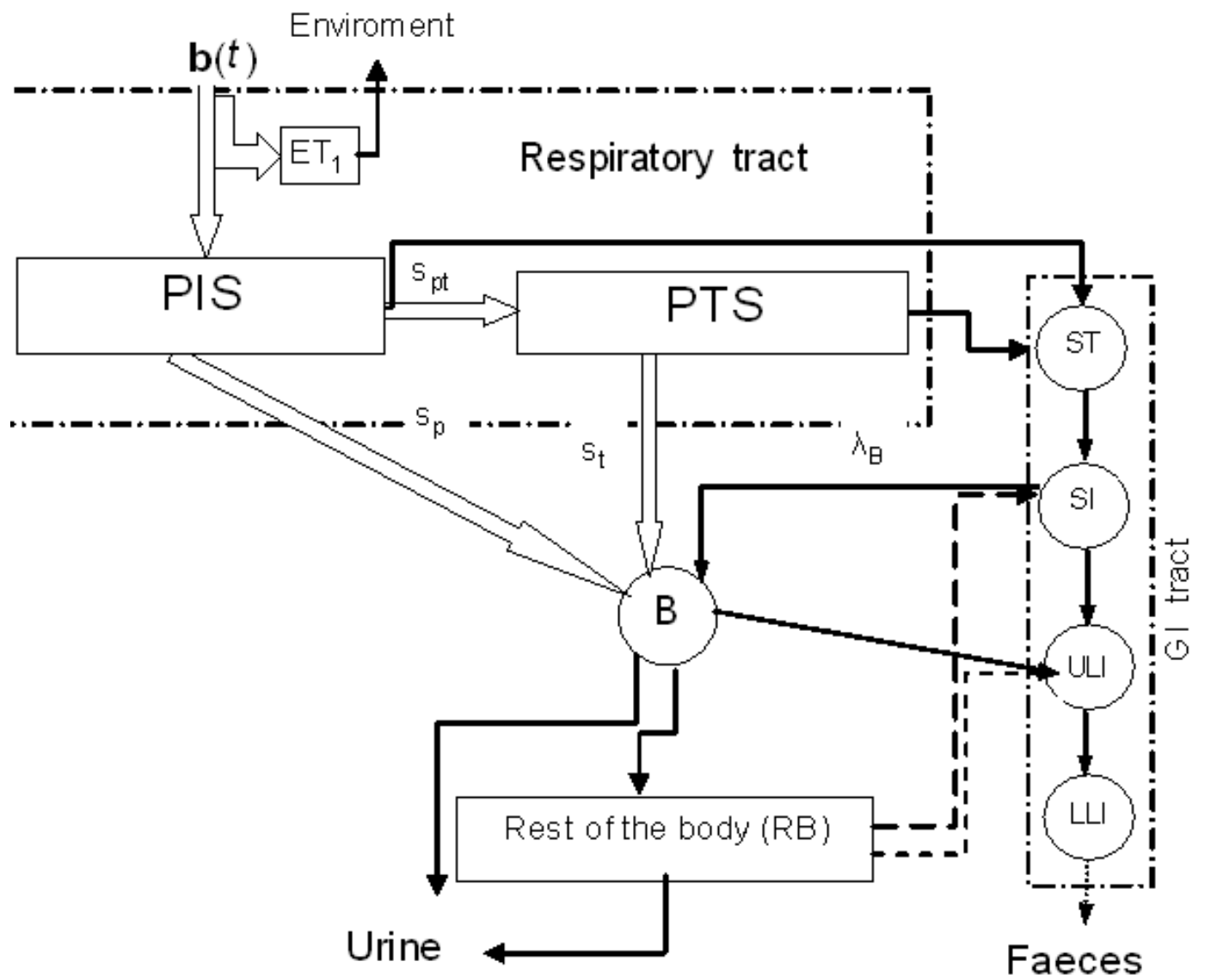
$$x_i(t) = \sum_{\gamma=1}^n a_{\gamma} e^{-k_{\gamma} t};$$

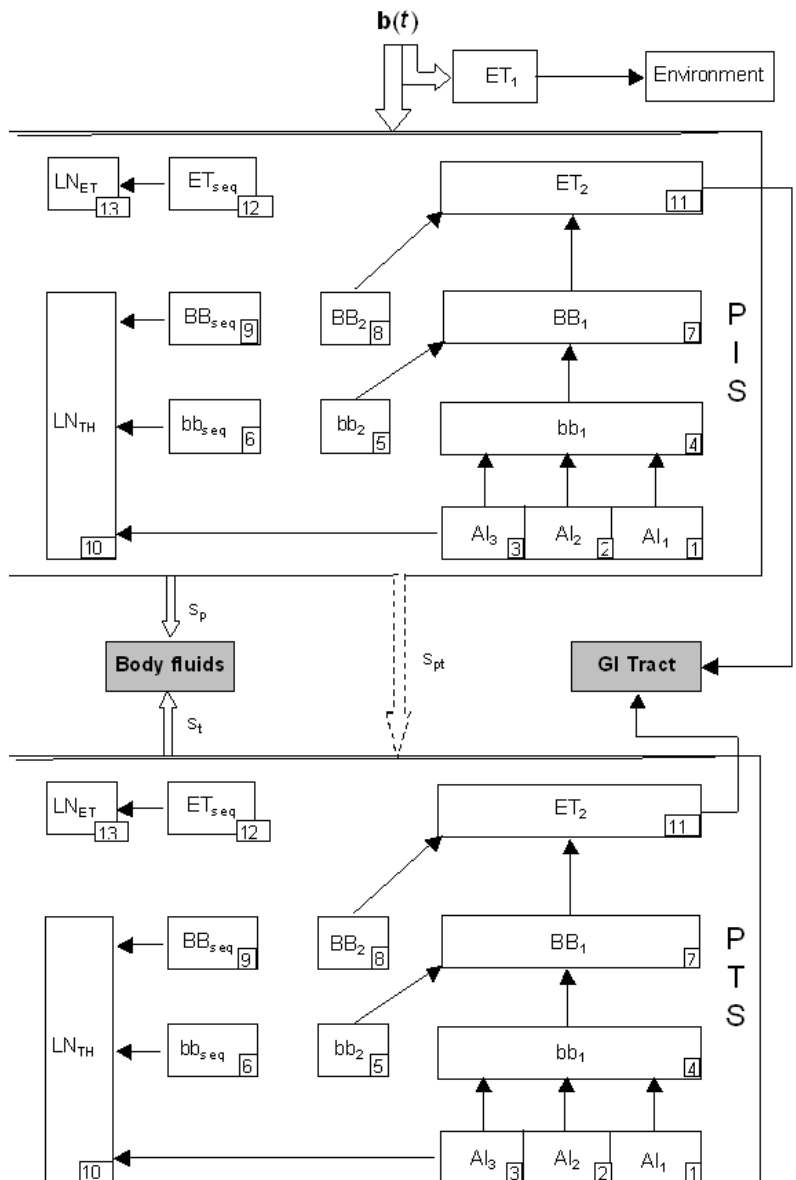
# Catenary branches



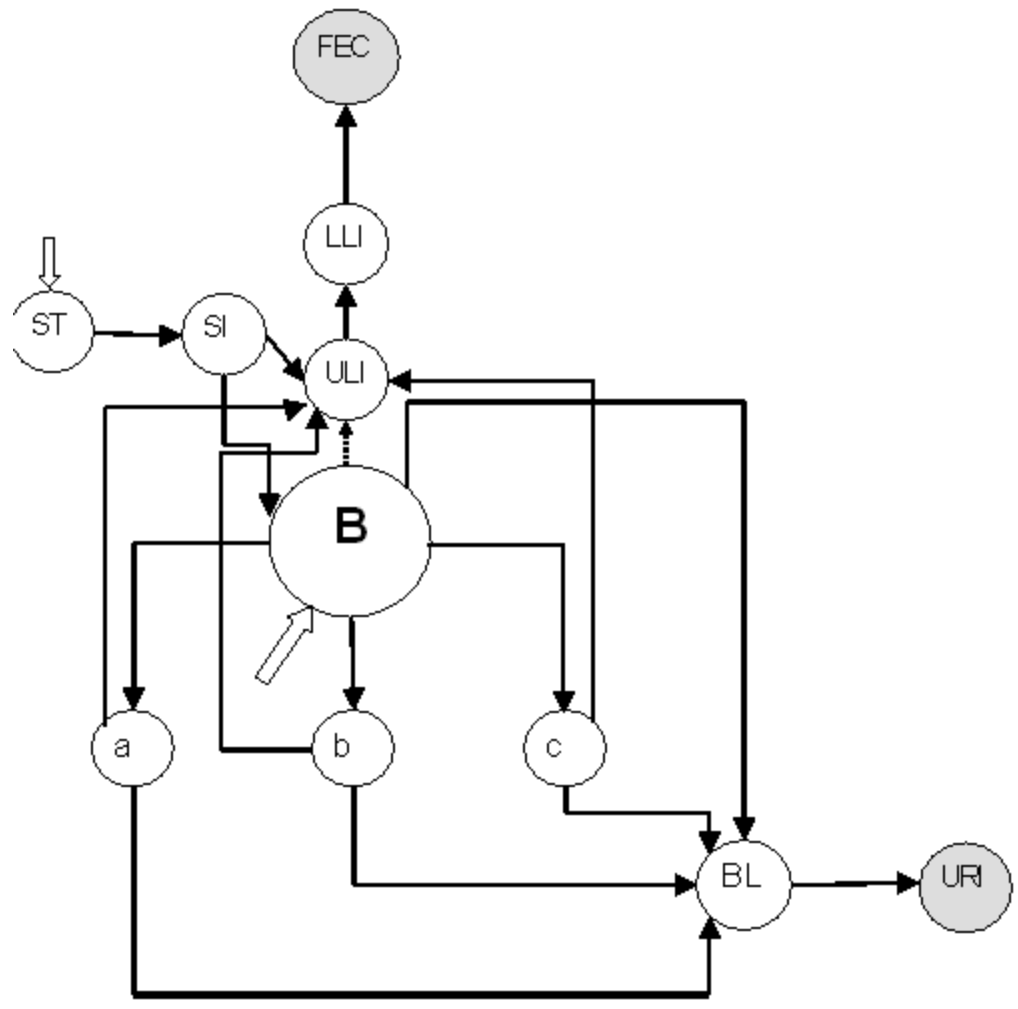
$$r_i(t) = e^{-\lambda_R t} \sum_C F_C \left( \prod_{p=1}^{i-1} k_p \right) \sum_{j=1}^i \left( \frac{e^{-K_j t}}{\prod_{\substack{p=1 \\ p \neq j}}^i (K_p - K_j)} \right) \quad i = 1, 2, \dots, n$$

Optimal Design applied to  
optimization of bioassays for  
people exposed to intake of  
radioactive substances





If a person intakes by inhalation instantaneously a quantity  $I$ , it is deposited directly in some compartments of PIS ( $Al$ ,  $bb_1$ ,  $bb_2$ ,  $bb_{seq}$ ,  $BB_1$ ,  $BB_2$ ,  $BB_{seq}$ ,  $ET_2$  and  $ET_{seq}$ ) and  $ET_1$ . The fraction deposited in each compartment is called Initial Deposition Factor or IDF. It is a function of Activity Median Aerodynamic Diameter (AMAD), which includes size, shape, density, anatomical and physiological parameters as well as various conditions of exposure. The . The general model of the RT is common to any element except the absorption rates  $\{s_{pt}, s_p, s_t\}$  that are related with the chemical form of the element. ICRP gives default values of absorption rates according to types F, M or S.





# Bioassay: Lung, urine and fecal excretion

$$y_i(t) = F_i(l_1, \dots, l_m, s_p, s_{pt}, s_t, f_1, \lambda_1, \dots, \lambda_n, h_1, \dots, h_r, \lambda_R, t) c_i$$

$$r_m(t) = \sum_{j,\nu} \text{IDF}_j(p) c_{j,\nu} e^{-d_{j,\nu} t}$$

$$\begin{aligned} & -0.000186053 e^{-4.35327 p - 2.0001 t} - 0.000533918 e^{-0.170111 p - 2.0001 t} - \\ & 0.0000805063 e^{-0.147244 p - 2.0001 t} + 0.0124915 e^{-0.0878945 p - 2.0001 t} + \\ & 0.0213271 e^{-4.35327 p - 0.0301 t} - 0.0110729 e^{-1.11147 p - 0.0301 t} + 0.0092284 e^{-0.147244 p - 0.0301 t} + \\ & 0.0110729 e^{-0.123578 p - 0.0301 t} + 0.0388835 e^{-0.170111 p - 0.0201 t} + \\ & 0.0768815 e^{-0.170111 p - 0.0011 t} + 0.0106723 e^{-0.170111 p - 0.00022 t} + \\ & 0.000148842 e^{-4.35327 p - 0.0001 t} - 0.00007751 e^{-1.11147 p - 0.0001 t} - \\ & 0.000120097 e^{-0.566783 p - 0.0001 t} + 0.00213432 e^{-0.170111 p - 0.0001 t} + \\ & 0.000064405 e^{-0.147244 p - 0.0001 t} + 0.00007751 e^{-0.123578 p - 0.0001 t} + \\ & 0.0000704456 e^{-0.0878945 p - 0.0001 t} + 0.000120097 e^{-0.0577835 p - 0.0001 t} \end{aligned}$$

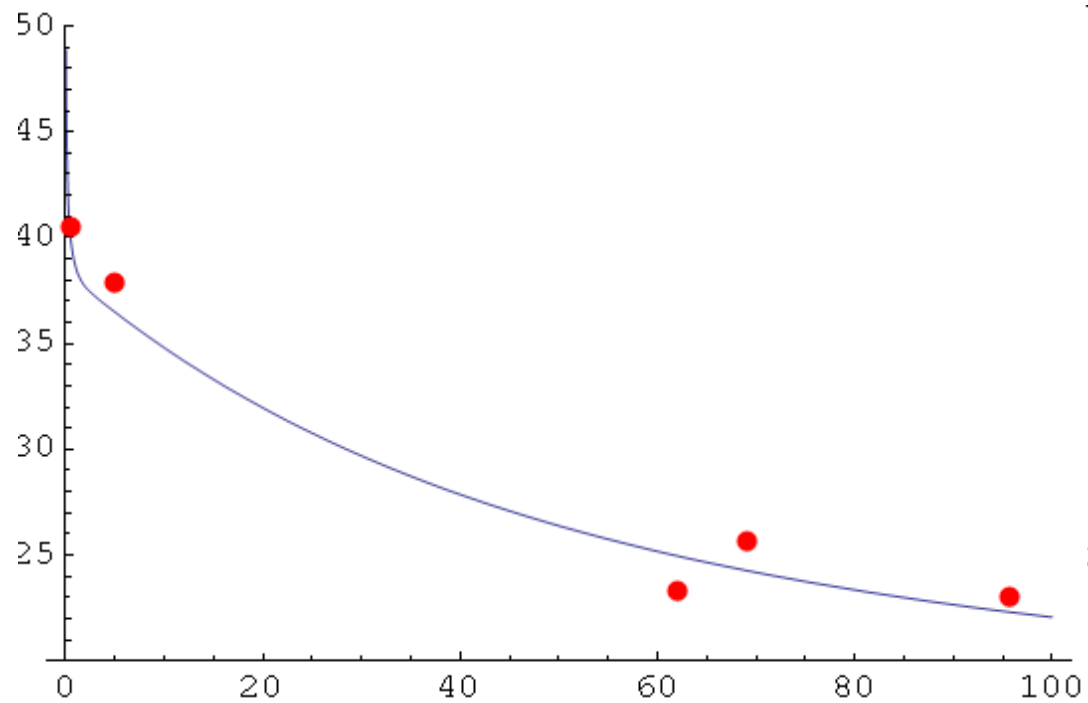
The below method is described in Sánchez G; Rodríguez-Díaz J. M. 2006

Let the function  $r(t, \beta)$  where  $\beta = \{\beta_1, \dots, \beta_p\}$  are parameters unknowns to be fitted using experimental data. We wish choose the best moments  $\{t_0, \dots, t_i, \dots, t_n\}$  to take the experimental data. It will be assumed that all measured have more or less the same uncertainties, that is  $\sigma^2 \simeq \sigma_i^2$ . The Fisher information matrix for a specific design  $\{t_1, \dots, t_i, \dots, t_n\}$  ( $t_i$  is the time when the  $i$ -th sample should be taken) will be used to compute the optimal design. A  $D$ -optimal design will be a design that leads the determinant of the information matrix to a maximum.

`Optdes [ R [ t,  $\beta_1, \dots, \beta_p$  ], t, {  $\{\beta_1, \beta_{10}\}, \dots, \{\beta_p, \beta_{p0}\}$  }, t0,  $\rho$ ,  $\sigma$ , n, opts ]` being  $\beta_1, \dots, \beta_p$  the unknown parameters and  $\beta_{10}, \dots, \beta_{p0}$  their the initial values, t<sub>0</sub> is the point were is takes the first measured,  $\rho$  (*relationship between samples that decays exponential*), usually will be used  $\rho=1$  (it is assumed that there not correlation between samples will be written "NoCorrelation");  $\sigma$  is the standard deviation of the measures, n is the number of the point (additional at t<sub>0</sub>) where we want to take measured; *opts* is a option to close the maximization method (The are the same that used by NMaximize).The function will give the values for  $\{t_0, \dots, t_i, \dots, t_n\}$ .

```
Table[{n, Optdes[inprlung[t, p], t, {{inp, 1000}, {p, 5}}, 0.5, 1, 2, n]}, {n, 1, 3}] //  
TableForm
```

```
1 0.000033437  
  t0 → 0.5 t1 → 70.2305  
2 0.0000668091  
  t0 → 0.5 t1 → 66.2152 t2 → 74.7861  
3 0.000100107  
  t0 → 0.5 t1 → 62.5845 t2 → 70.6721 t3 → 78.8042
```





## Optimal design

If we wish to fit the expression  $R[t, b_1, \dots, b_p]$  being  $b_1, \dots, b_p$  the unknown parameters to be fitted and we want take  $n$  experimental data the program gives the best moments  $t_i$  where the  $n$  samples should be taken. The user must introduce as input:  $R[t, b_1, \dots, b_p]$ ,  $t_0$  (the point where the first sample is taken);  $std$  (standard deviation of the measures)  $n$  (number of the points to take the sample, additional at  $t_0$ ),  $\rho$  (correlation between measures, 1 by default). [Help](#)

A detail description of this program can be found in: G.Sanchez; J.M.Rodríguez-Díaz. Optimal design and mathematical model applied to establish bioassay programs: Radiation Protection Dosimetry. doi:10.1093/rpd/nci499.2007. ISSN/ISBN:ISSN 1742-3406;

Expression:  Variable   $t_0$  (point for first sample):

Parameters and starting points:  Rho (Correlation):  std(standard deviation):  Number of points where to take the samples:

Evaluate

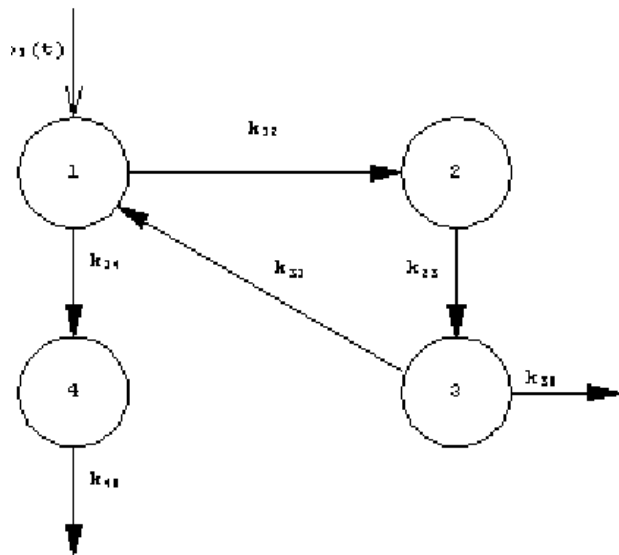
*results*  
 (0.16647,  $\{t_0 \rightarrow 0.5, t_1 \rightarrow 1.98703, t_2 \rightarrow 8.16929, t_3 \rightarrow 15.0829\}$ )

# Futures Development: multiresponse models

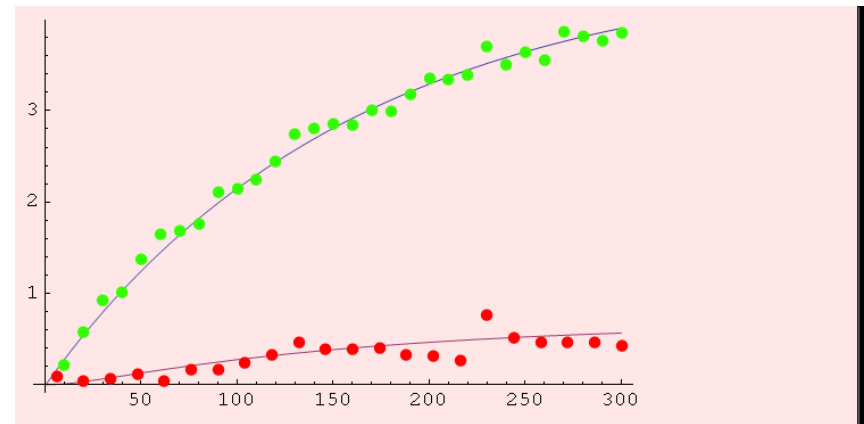
Suppose that  $k_{12}$  and  $k_{23}$  are unknown. We want estimated  $k_{12}$  and  $k_{23}$  using experimental data of  $x_2(t)$  and  $x_3(t)$ .

The usual way is fitting  $x_1 = f_1(t, k_{12}, k_{23})$  and  $x_2 = f_2(t, k_{12}, k_{23})$  but the analytical expression can not be obtained.

$$\begin{pmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \end{pmatrix} = \begin{pmatrix} -(k_{10} + k_{12} + k_{13}) & k_{21} & k_{31} \\ k_{12} & -(k_{20} + k_{21} + k_{23}) & k_{32} \\ k_{13} & k_{23} & -(k_{30} + k_{31} + k_{32}) \end{pmatrix} \begin{pmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \end{pmatrix}$$



0.508007, | k12 = 0.830821, k23 = 0.00855833 |



Could be applied Optimal Design for this situation (Multiresponse problem where the analytical expression of the functions to be fitted can not be obtained however the underline SODE is known)?