Accuracy of transient methods

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Concept

●Experiment	$\{t_n,T_n\}_{n=1}^N$
	$\mathbf{b} = \{b_j\}_{j=1}^{N_b}$
●Model	$T_{model} = f(t, \mathbf{a}, \mathbf{b})$
	$\mathbf{a}=\{a_i\}_{i=1}^{N_a}$

•Fitting model parameters to experimental data:

determined, overdetermined problem

Accuracy estimation - overdetermined problem

Least squares optimization

$$\min\{\sum_{n=1}^{N}[T_n - f_n(\mathbf{a}, \mathbf{b})]^2\}$$

where $f_n(\mathbf{a},\mathbf{b})=f(t_n,\mathbf{a},\mathbf{b})$

 $egin{array}{l} t_n \ \mathbf{b}_n \ \{T_n\}_{n=1}^N \ \mathbf{a} \end{array}$

... deterministic parameter

- ... independent random variables
- ... dependent random variables

$$\sum_{n=1}^{N} (T_n - f_n) \frac{\partial f_n}{\partial a_i} = 0, \qquad i = 1, 2, \dots N_a$$
(1)

There are estimated the moments:

Now we estimate the uncertainty

$$u(a_i)^2 \sim < (a_i - < a_i >)^2 > = < (\Delta a_i)^2 > \sim < (da_i)^2 >$$
 (6)

Differentiating the equation (1) we obtain

$$\sum_{i=1}^{N_a} A_{ki} da_i = \sum_{n=1}^{N} dT_n \frac{\partial f_n}{\partial a_k} + \sum_{j=1}^{N_b} B_{kj} db_j, \quad k = 1, 2, \dots N_a$$
(7)

where

$$A_{ki} = \sum_{n=1}^{N} \left[(f_n - T_n) \frac{\partial^2 f_n}{\partial a_k \partial a_i} + \frac{\partial f_n}{\partial a_k} \frac{\partial f_n}{\partial a_i} \right] \simeq \sum_{n=1}^{N} \frac{\partial f_n}{\partial a_k} \frac{\partial f_n}{\partial a_i}$$
(8)
$$B_{kj} = -\sum_{n=1}^{N} \left[(f_n - T_n) \frac{\partial^2 f_n}{\partial a_k \partial b_j} + \frac{\partial f_n}{\partial a_k} \frac{\partial f_n}{\partial b_j} \right] \simeq -\sum_{n=1}^{N} \frac{\partial f_n}{\partial a_k} \frac{\partial f_n}{\partial b_j}$$
(9)

Solution of equation (7) has the form

$$da_{k} = \sum_{i=1}^{N_{a}} A_{ki}^{-1} \Big(\sum_{n=1}^{N} dT_{n} \frac{\partial f_{n}}{\partial a_{i}} + \sum_{j=1}^{N_{b}} B_{ij} db_{j} \Big), \quad k = 1, 2, \dots N_{a}$$
(10)

and the uncertainty of a_k is with respect to equation (2) - (6) and (10)

$$u(a_k)^2 = \sum_{i=1}^{N_a} \sum_{i'=1}^{N_a} A_{ki}^{-1} A_{ki'}^{-1} \Big[A_{i'i} u(T)^2 + \sum_{j=1}^{N_b} \sum_{j=1}^{N_b} B_{ij} B_{i'j} u(b_j)^2 \Big]$$
(11)

$$u(a_k)^2 = C_{kT}^2 u(T)^2 + \sum_{j=1}^{N_b} C_{kj}^2 u(b_j)^2$$

where the sensitivities are

$$C_{kT} = \sqrt{A_{kk}^{-1}} \tag{12}$$

and

$$C_{kj} = \sum_{i=1}^{N_a} A_{ki}^{-1} B_{ij}$$
(13)

The elements of matrix A and B are defined with equations (8) and (9). It is seen that

$$\mathbf{A} \sim N_{_{7}} \mathbf{B} \sim N$$

therefore $C_{kT} \sim \frac{1}{\sqrt{N}}$ and C_{kj} is N-independent. For power-like dependence it is useful to define indices

$$\nu_{kj} = \frac{\partial \log(a_k)}{\partial \log(b_j)} = \frac{b_j}{a_k} C_{kj}$$
(14)

Then for relative uncertainties we can write the equation

$$u_r(a_k)^2 = C_{kT}^2 \frac{u(T)^2}{a_k^2} + \sum_{j=1}^{N_b} \nu_{kj}^2 u_r(b_j)^2$$
(15)

Application to simple model

1D model: The temperature response on the stepwise heat flow from planar heat source: 0

$$T(t,x) = T_0 \left[\frac{e^{-v^2}}{\sqrt{\pi}v} - \Phi^*(v) \right]$$
(16)

where $T_0 = qx/\lambda$

- q ... heat flow density at source
- x ... axial space coordinate of thermometer
- λ ... thermal conductivity

$$v = \frac{x}{2\sqrt{k}}$$

 $v = \frac{\omega}{2\sqrt{kt}}$ k ... thermal diffusivity

t time

We apply the formula (13) on the relation (16). The obtained indices are constant ones:

$$\begin{array}{c|c} \nu_{k,x} = 2 & \nu_{k,q} = 0 \\ \hline \nu_{\lambda,x} = 1 & \nu_{\lambda,q} = 1 \\ _{9} \end{array}$$

Application to more complicated model

2D model: The temperature response on the stepwise heat flow from planar heat source - cylindric sample with radial Newtonian heat transfer from sample surface:

$$T(t,x) = T_0 \frac{R}{x} \sum_{\xi} \frac{\beta}{\xi(\xi^2 + \beta^2) J_0(\xi)} \left[e^{-2uv} \Phi^*(v-u) - e^{2uv} \Phi^*(v+u) \right]$$
(17)

where Φ^* is the complementary error function,

$$u = \xi \frac{1}{R}$$

 $R \dots$ sample radius

$$\beta = \frac{R\alpha}{\lambda}$$

lpha ... heat transfer coefficient

 ξ is the root of the equation

$$\beta J_0(\xi) - \xi J_1(\xi) = 0$$

 J_0, J_1 ... Bessel functions of the first kind

Diffusivity indices



Figure 1:



Conductivity indices

Figure 2:



Indices of heat transfer coefficient

Figure 3: