

# Nonlinear modeling: Case study: river discharge

## Introduction to Statistical Modelling

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## Model overview



## Streeter-Phelps model

Use water pollution as water quality monitoring tool. Describes how dissolved oxygen decreases in a river along a certain distance by degradation of biological oxygen demand.

- Aerobic bacteria gradually remove organic pollution downstream of pollution source
- Reactions
  - Aerobic removal of biochemical oxygen demand
  - Oxygen transfer between atmosphere and water
- Assumption: plug-flow stream
- Simple dynamical model (nonlinear)

# Setting

## Typical values for rivers

- Biochemical oxygen demand (BOD)
  - Not polluted:  $\text{BOD} < 1\text{mg/l}$
  - Mildly polluted:  $2\text{mg/l} < \text{BOD} < 8\text{mg/l}$
- Dissolved oxygen (DO)
  - Maximal saturation:  $\text{DO} = 12.9\text{mg/l}$
  - Typical value in freshwater stream:  $\text{DO} \approx 9\text{mg/l}$
  - Threat to aquatic life:  $\text{DO} < 5\text{mg/l}$

# Model

Constant flow rate: location doesn't matter, dynamic model **in time**

$$\begin{aligned}\frac{dBOD}{dt} &= BOD_{in} - k_1 BOD \\ \frac{dDO}{dt} &= k_2 (DO_{sat} - DO) - k_1 BOD\end{aligned}$$

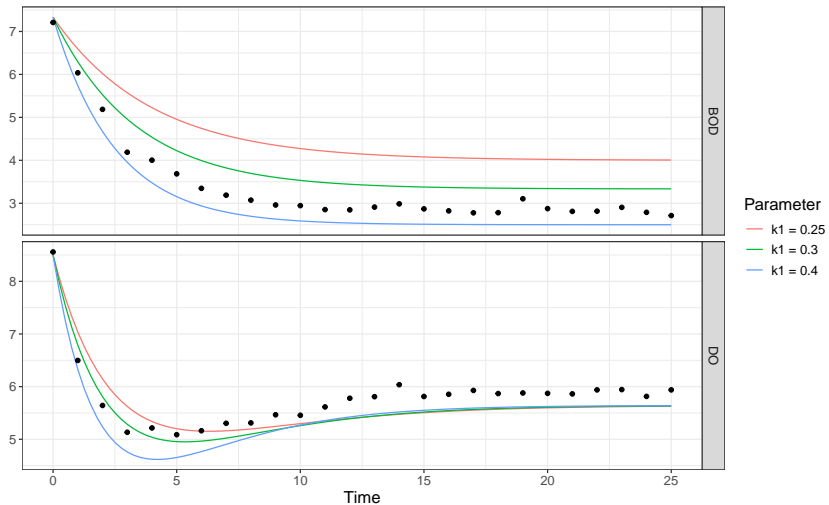
where

- $BOD_{in}$ : BOD flux of waste discharge ( $\text{mg} \cdot \text{l}^{-1} \cdot \text{min}^{-1}$ )
- $DO_{sat}$ : dissolved oxygen concentration at saturation
- $k_1$ : deoxygenation rate ( $\text{min}^{-1}$ )
- $k_2$ : reaeration rate, rate at which oxygen can be absorbed from the atmosphere ( $\text{min}^{-1}$ )

# Model inputs

- Initial conditions:
  - $\text{BOD}_{t=0} = 7.33\text{mg/l}$
  - $\text{DO}_{t=0} = 8.5\text{mg/l}$
- Initial model inputs:
  - $\text{BOD}_{in} = 1\text{mg}\cdot\text{l}^{-1} \cdot \text{min}^{-1}$
  - $\text{DO}_{sat} = 8.5\text{mg}\cdot\text{l}^{-1}$
  - $k_1 = 0.3\text{min}^{-1}$  (**unknown**)
  - $k_2 = 0.4\text{min}^{-1}$  (**unknown**)

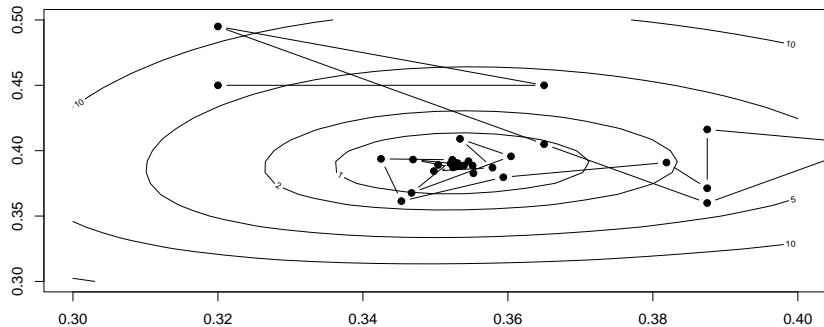
# Model trajectories



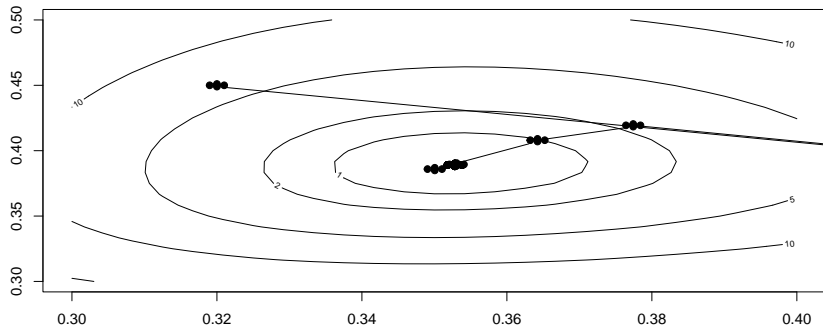
## Parameter estimation



# Parameter estimation - simplex method

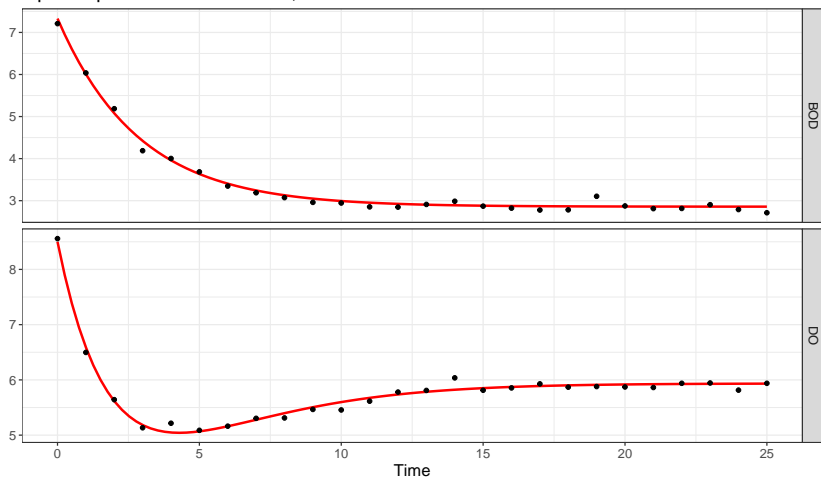


## Parameter estimation - BFGS



# Optimal parameter result

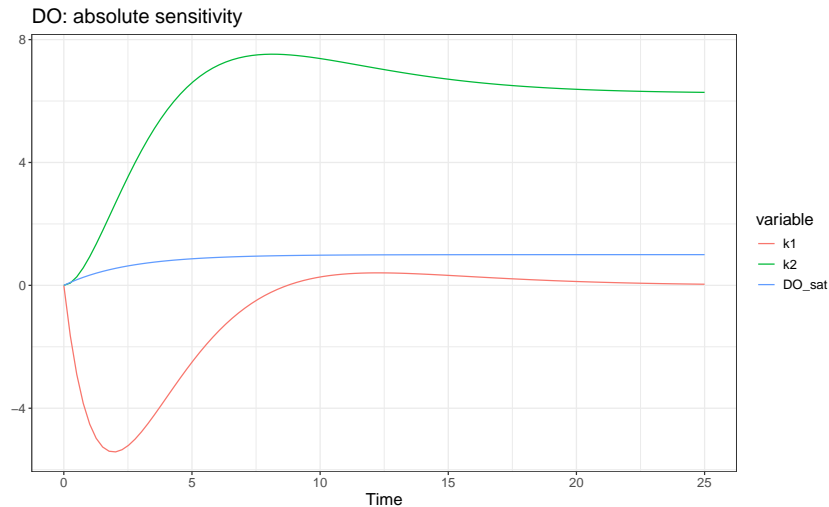
Optimal parameters:  $k_1 = 0.35$ ,  $k_2 = 0.39$



## Sensitivity analysis

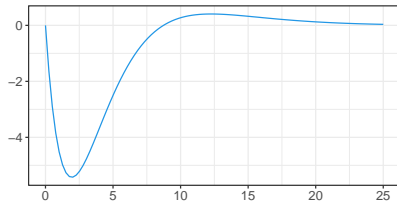
# Absolute sensitivity functions

Absolute sensitivity of DO with respect to  $k_1$  and  $k_2$ .

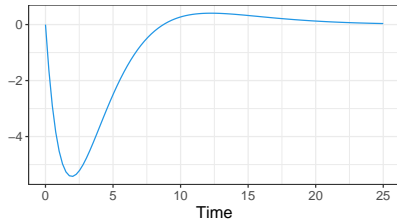


# Difference exact approximate

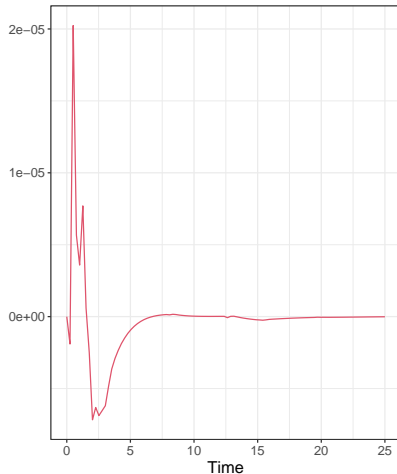
Exact sensitivity



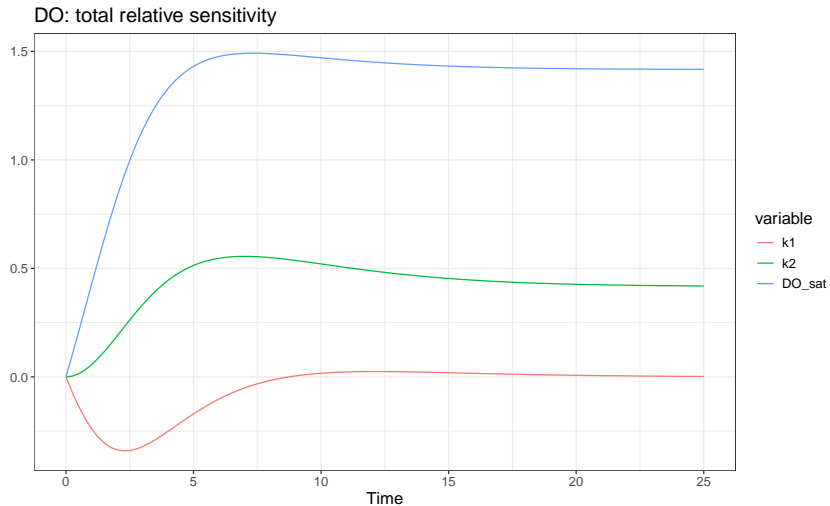
Approximate sensitivity



Difference exact – approximate



# Relative sensitivity functions



# Conclusions

- Note difference in values with absolute sensitivities
- DO seems more sensitive to  $k_2$  than to  $k_1$
- Extrema at slightly different time points: information concerning correlation between both parameters
  - Maximal sensitivity at same time point: parameters strongly correlated (impact of change in parameters similar)
- Studying sensitivity very valuable: sensitivity used in many techniques for model analysis



## Aside: quality of estimation

- DO measurements at different times: measurement error (obtained manually) is  $0.05 \text{ mg}\cdot\text{l}^{-1}$
- Estimate simultaneously  $k_1$  and  $k_2$  (assume all other parameters and initial conditions constant)
- Gives  $k_1 = 0.353\text{min}^{-1}$ ,  $k_2 = 0.389$

# Quality of estimation: FIM

Fisher information matrix:

$$\text{FIM} = \sum_{i=1}^N \left( \frac{\partial y}{\partial \theta}(t_i) \right)^T Q_i \left( \frac{\partial y}{\partial \theta}(t_i) \right)$$

- Measurement noise  $\sigma_{DO} = 0.05 \text{mg l}^{-1}$
- Weight “matrix” in the objective function  $Q = \sigma_{DO}^{-2}$ .

Gives:

$$\text{FIM} = \begin{bmatrix} 3.91 \cdot 10^4 & -2.96 \cdot 10^4 \\ -2.96 \cdot 10^4 & 4.56 \cdot 10^5 \end{bmatrix}$$

## Quality of estimation: confidence intervals

Error covariance matrix:

$$C = \text{FIM}^{-1} = \begin{bmatrix} 2.69 \cdot 10^{-5} & 1.75 \cdot 10^{-6} \\ 1.75 \cdot 10^{-6} & 2.31 \cdot 10^{-6} \end{bmatrix}$$

95% confidence intervals:

$$k_1 : 0.353 \pm 0.011$$

$$k_2 : 0.389 \pm 0.003$$

Covariance:

$$\text{cor}(k_1, k_2) = 0.22$$

# Quantitative analysis

- Calculate  $\delta_{rmsq}$  for DO and different parameters
- Illustration: different measuring schemes for DO:
  - Scheme 1:  $t_k = 0 : 0.1 : 2$
  - Scheme 2:  $t_k = 0 : 2 : 20$
  - Scheme 3:  $t_k = 0 : 2 : 10$
  - Scheme 4:  $t_k = 10 : 2 : 20$

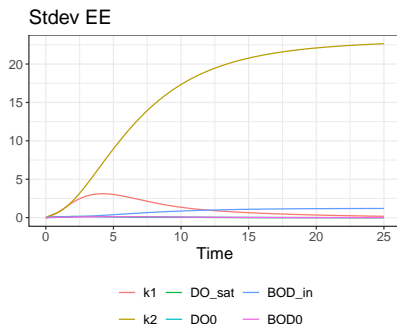
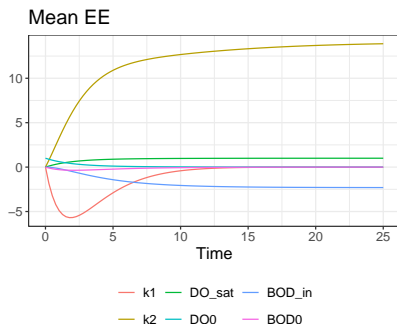
	Scheme 1	Scheme 2	Scheme 3	Scheme 4
$k_1$	0.24	0.13	0.18	0.18
$k_2$	0.09	0.44	0.43	0.43
$DO_{sat}$	0.49	1.32	1.23	1.23
	$k_2 < k_1 < DO_{sat}$	$k_1 < k_2 < DO_{sat}$	$k_1 < k_2 < DO_{sat}$	$k_1 < k_2 < DO_{sat}$

## GSA: Morris screening: parameter ranges

- Global sensitivity for 6 parameters and initial conditions in designated ranges:

$$\begin{array}{lll} k_1 : [0.1; 0.6] & k_2 : [0.1; 0.6] & \text{DO}_{sat} : [10; 12] \\ \text{BOD}_{in} : [0.1; 2] & \text{DO}_{t=0} : [6; 10] & \text{BOD}_{t=0} : [6; 10] \end{array}$$

# GSA: Morris screening

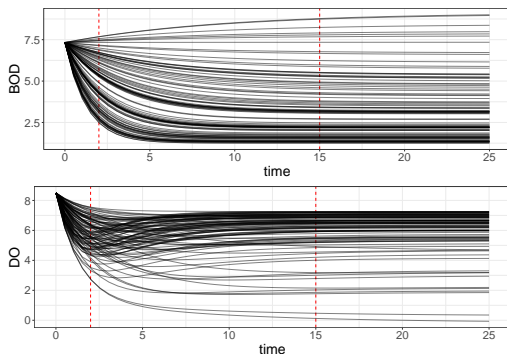


Among the selected parameters:

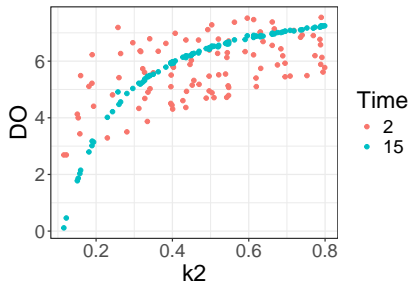
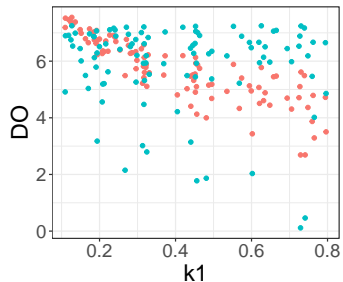
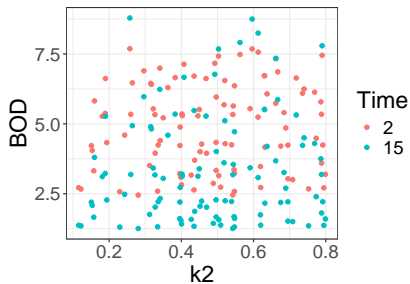
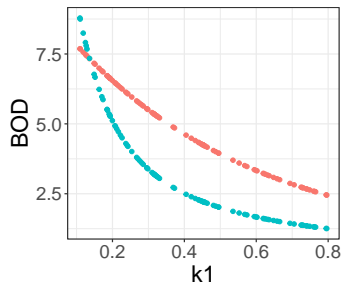
- At time 0, DO is only sensitive to  $DO_0$
- At time 25, DO is sensitive to  $k_2$  and  $DO_{sat}$ .

# GSA: Monte Carlo

- Monte Carlo: 100 simulations with varying  $k_1$  and  $k_2$
- Parameter ranges:  $k_1, k_2$  uniformly sampled from  $[0.1, 0.8]$
- Interested in parameter effect at  $t = 2$  and  $t = 15$



# GSA: Monte Carlo





## GSA: Standardized regression coefficients

Previous plots show:

- *BOD* is sensitive to  $k_1$  at  $t = 2, 15$
- *DO* is sensitive to  $k_2$  at  $t = 15$  (and somewhat at  $t = 2$ )

Regression coefficients for *DO* at  $t = 15$ :

$$\begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} -0.342 \\ 6.374 \end{bmatrix}.$$

Standardized regression coefficients (using  $\sigma_{k_1} = 0.211$  and  $\sigma_{DO} = 1.887$ ):

$$SRC_{k_1} = -0.038, \quad \text{and} \quad SRC_{k_2} = 0.713.$$