



MOSAIC_{bioacc} REPORT

2021-04-15

This report is provided by the $MOSAIC_{bioacc}$ application available here: https://mosaic.univ-lyon1.fr/bioacc

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MOSAIC_{bioacc} uses the JAGS (version 4.3.0) and R (version 4.0.2) software, and in particular packages RJags (version 4.10), jagsUI (version 1.5.1) and Shiny (version 1.6.0).

The MOSAIC_{bioacc} application is a turn-key web tool providing bioaccumulation factors (BCF/BSAF/BMF) from a toxicokinetic (TK) model fitted to accumulation-depuration data. It is designed to fulfil the requirements of regulators when examining applications for market authorization of active substances.

Data summary

File used: Gammarus dichlorophenol 0.979d Ashauer2012.txt

Exposure: 1651.678 $\mu g.mL^{-1}$

Accumulation phase duration: 0.979 days

Number of replicates: 2

 $\text{Times: } 0,\, 0.207,\, 0.208,\, 0.541,\, 0.542,\, 0.978,\, 0.979,\, 1.208,\, 1.457,\, 1.458,\, 1.999,\, 2,\, 2.999,\, 3,\, 3.999,\, 4,\, 5.02,\, 5.021,\, 3$

5.999, 6

Exposure routes: water

Elimination routes: excretion biotransformation

Bayesian inference

Three MCMC chains were used to estimate model parameters.

Number of iterations: 146094

Thin: 39





TK Model

The TK model used for these calculations was:

$$\frac{dC_p(t)}{dt} = k_{uw} \times c_w - (k_{ee} + k_{m1} + k_{m2}) \times C_p(t) \quad \text{for } 0 \le t \le t_c$$

$$\frac{dC_p(t)}{dt} = -(k_{ee} + k_{m1} + k_{m2}) \times C_p(t) \quad \text{for } t > t_c$$

$$\frac{dC_{m1}(t)}{dt} = k_{m1} \times C_p(t) - k_{em1} \times C_{m1}(t)$$

$$\frac{dC_{m2}(t)}{dt} = k_{m2} \times C_p(t) - k_{em2} \times C_{m2}(t)$$

with:

t: time (expressed in days)

 t_c : duration of the accumulation phase (expressed in days)

 $C_p(t)$: internal concentration of the parent compound at time (expressed in $\mu g.g^{-1}$)

 k_{ee} : elimination rates of excretion (expressed per days $^{-1}$)

 c_w : exposure concentration of water route (expressed in $\mu g.mL^{-1}$)

 k_{uw} : uptake rate of water exposure (expressed per days $^{-1}$)

 $C_{m\ell}(t)$: internal concentration of metabolite ℓ (expressed in $\mu g.g^{-1}$)

 ℓ : index of metabolites, $\ell=1$... L with L total number of metabolites

 $k_{m\ell}$: metabolization rate of metabolite ℓ (expressed per days $^{-1}$)

 $k_{em\ell}$: elimination rates of metabolite ℓ (expressed per days $^{-1})$

Bioaccumulation factor calculation

Calculations

$$BCF_k = \frac{k_{uw}}{k_{ee} + k_{m1} + k_{m2}}$$

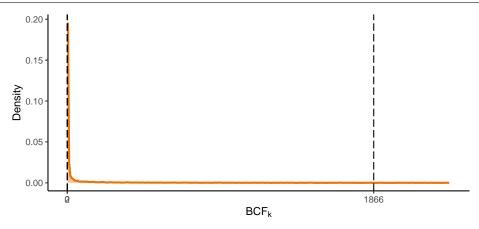
$$BCF_{ss} = \frac{C_p(t_c)}{c_w}$$

Bioconcentration factor (BCF)

BCF_k plot





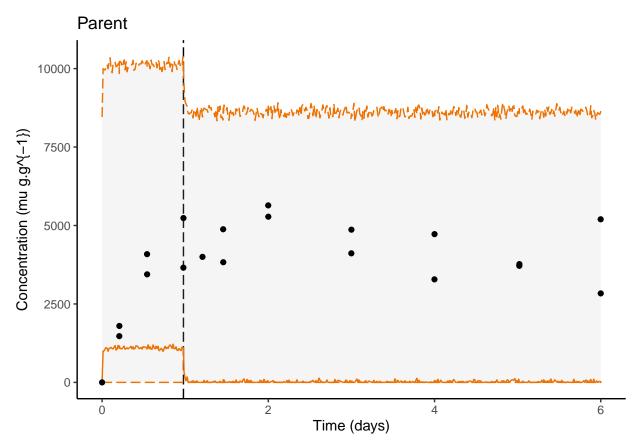


BCF summary

	2.5%	50%	97.5%	CV
BCFk	0	2	1866	230

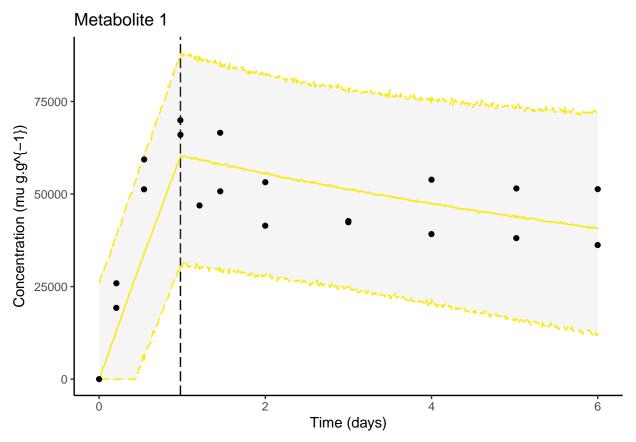
Fitting results

Fit plot



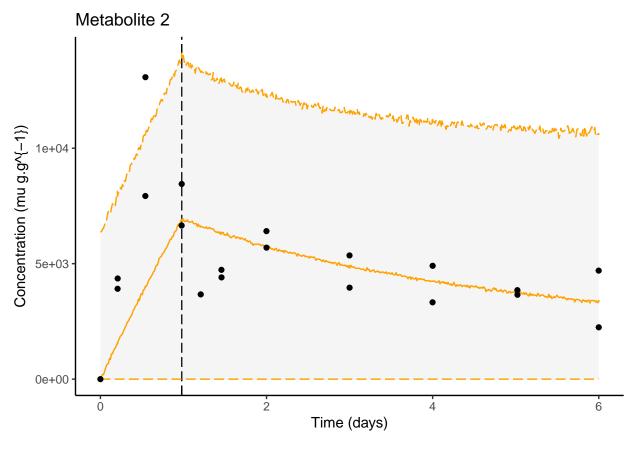












Quantiles of estimated parameters

	2.5%	50%	97.5%	
$\overline{k_{uw}}$	35.52	51.39	56470	d^{-1}
k_{ee}	2.208e-05	49.87	51870	d^{-1}
k_{m1}	12.82	185.5	73830	d^{-1}
k_{m2}	1.322	21.84	8819	d^{-1}
k_{em1}	3.724e-05	0.08068	0.1752	d^{-1}
k_{em2}	2.109e-05	0.1431	0.6486	d^{-1}
σ_p	3047	4167	6144	$\mu g.g^{-1}$
σ_{met1}	8795	12460	18990	$\mu g.g^{-1}$
σ_{met2}	2201	3068	4597	$\mu g.g^{-1}$

Goodness-of-fit criteria

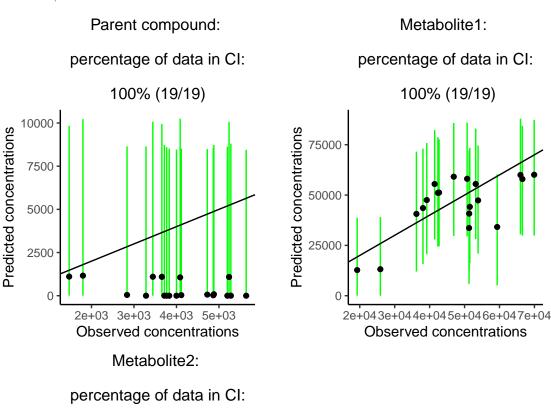
Posterior Predictive Check

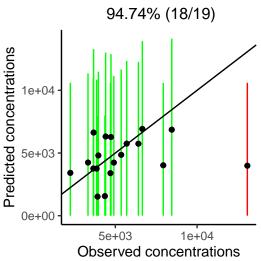
The PPC shows the observed values against their corresponding estimated predictions (black dots), along with their 95% credible interval (vertical segments). If the fit is correct, we expect to see 95% of the data within the intervals. Ideally observations and predictions should coincide, so we would expect to see black





dots along the first bisector y = x (plain black line). The 95% credible intervals are colored in green if they overlap this line, in red otherwise.



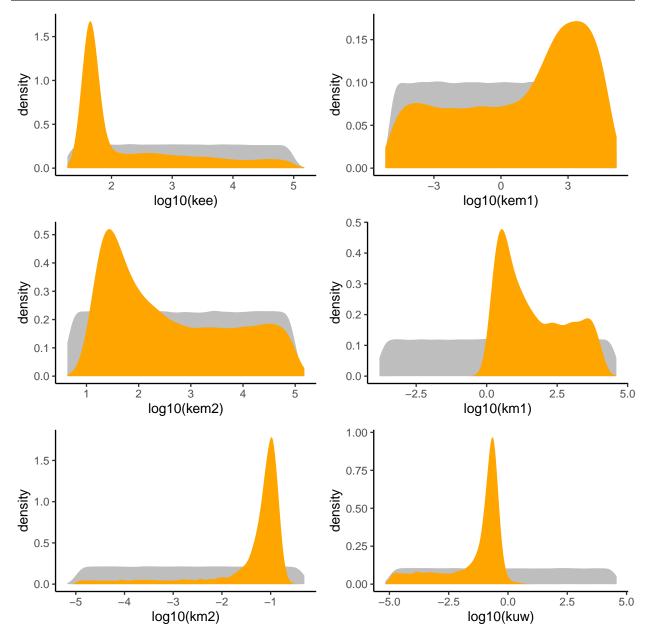


Priors and posteriors

The prior distribution is represented by the gray area and the posterior distribution by the orange area. The accuracy of the model parameter estimation can be visualized by comparing prior and posterior distributions: the overall expectation is to get a narrower posterior distribution compared to the prior one, what reflects that data contributed enough to precisely estimate parameters.

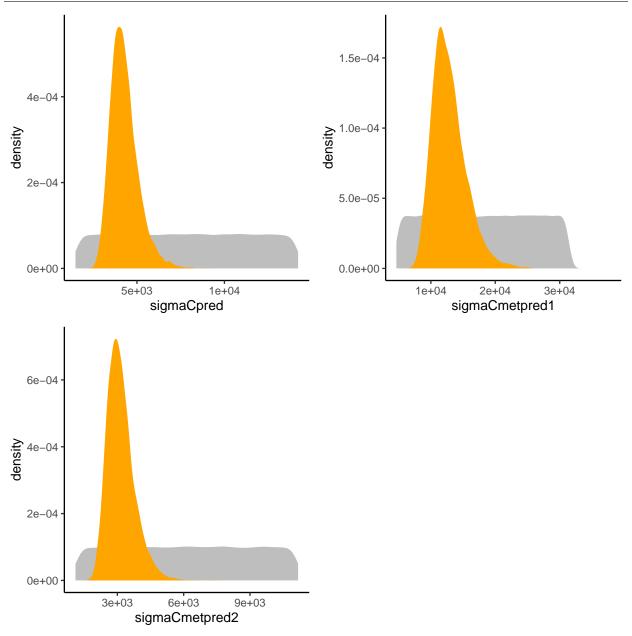












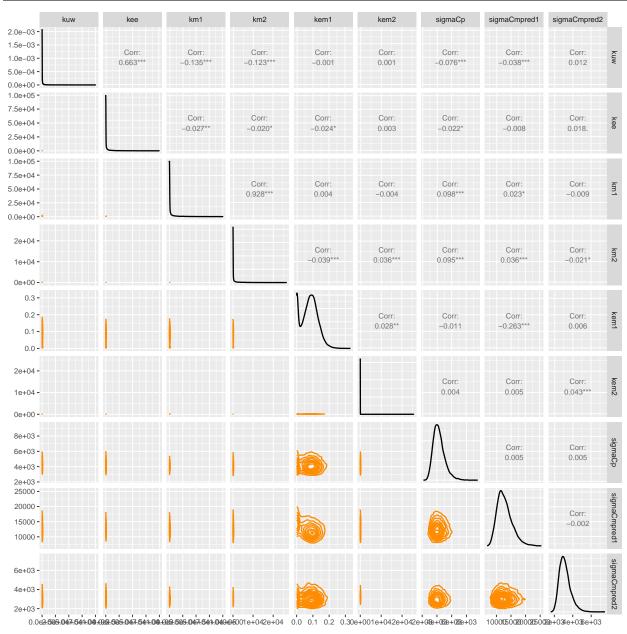
Correlation between parameters

If you want to see the coloured matrix giving a summary of parameter correlations, you need to import the corresponding figure directly from the application, page bottom, section "Downloads", then choose Download an output and select "GOF" then "parameter correlation". You can select the output format you prefer.

Correlations between parameters are visualized by projecting the joint posterior distribution in a plot matrix with planes of parameter pairs (lower triangular elements), marginal posterior distribution of each model parameter (diagonal), and Pearson correlation coefficients (upper triangular elements). Correlations are expected to be low (reflected by "potatoid" shapes of density lines in orange); a leaning elliptical shape translates high correlations (positive if leaning to the right, negative if leaning to the left).







Potential Scale Reduction Factors

Convergence of the MCMC chains can be check with the Gelman-Rubin diagnostic expressed with the potential scale reduction factor (PSRF). Approximate convergence is diagnosed when the PSRF is below 1.01.

	PSRF
kuw	1.098
kee	1.062
km1	1.003
km2	1.006
kem1	1.001
kem2	1.283
sigmaCpred	1
${\rm sigmaCmetpred1}$	1.019





	PSRF
$\overline{{\rm sigmaCmetpred2}}$	1

Watanabe-Akaike information criterion

Information criteria offer a computationally appealing way of estimating the generalization performance of the model. A fully Bayesian criterion is the widely applicable information criterion (WAIC) by Watanabe a penalized deviance statistics accounting for the uncertainty in the parameters and can be used also for singular models. WAIC is widely used in model comparison for a same dataset (e.g., with or without $k_{\rm ee}$). Sub-models with lower WAIC values will be preferred.

WAIC = 1147

Deviance Information Criterion

This criteria, denoted DIC, is a penalized deviance statistics accounting for the number of parameters for use in model comparison for a same dataset (e.g., with or without $k_{\rm ee}$). Sub-models with lower DIC values will be preferred.

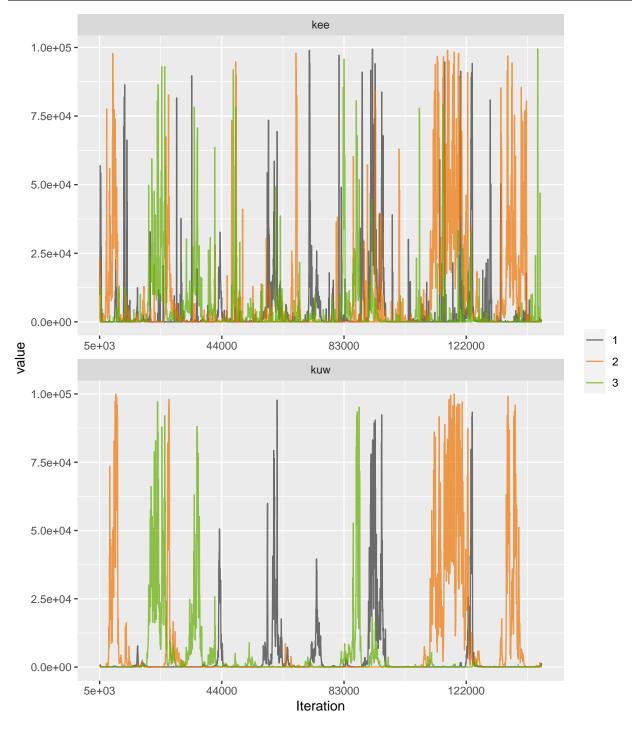
DIC = 1155

Traces of MCMC iterations

A traceplot is an essential plot for assessing convergence and diagnosing of MCMC chains. It shows the time series of the sampling process leading to the posterior distribution. Different colors are used for each of the chains (here 3) to assess within-chain convergence.

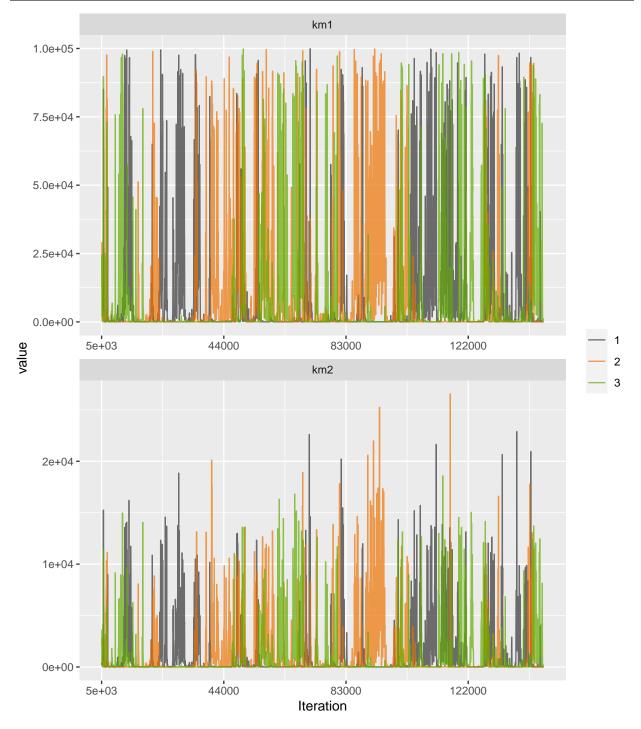






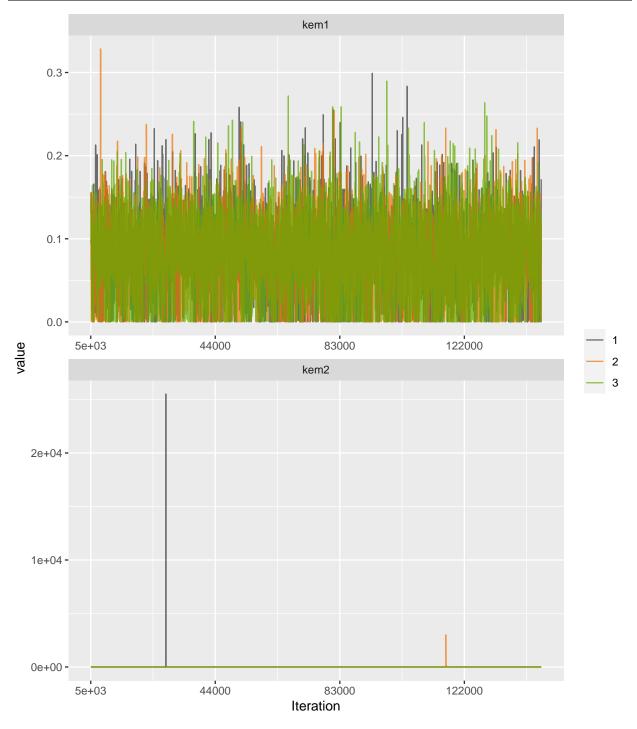






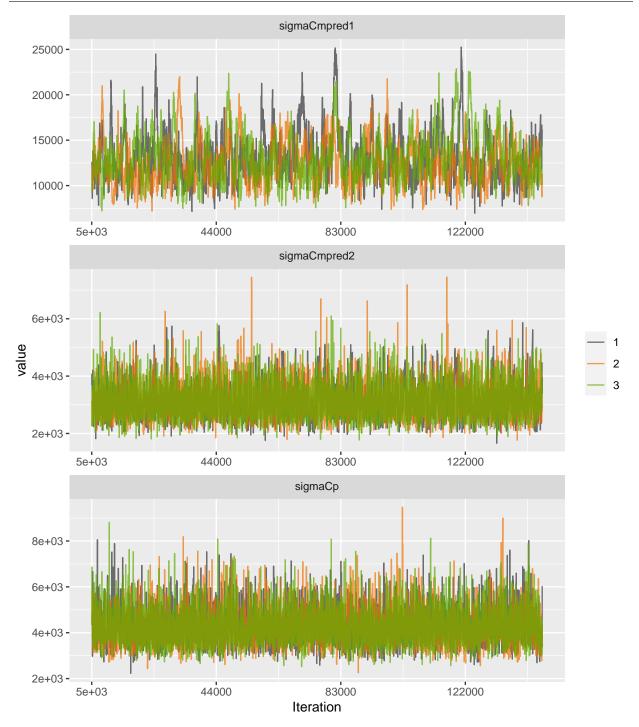
















Data Table

conc	concm1	expw	replicate	concm2
0	0	1652	1	0
1474	19259	1652	1	3913
1799	25897	1652	2	4360
4087	51282	1652	1	7931
3443	59337	1652	2	13076
3656	69927	1652	1	6651
5237	66001	1652	2	8452
4000	46909	1652	2	3670
4880	66550	1652	1	4733
3829	50732	1652	2	4406
5640	41439	1652	1	6410
5280	53192	1652	2	5695
4113	42367	1652	1	5357
4866	42697	1652	2	3961
3283	39179	1652	1	3328
4725	53867	1652	2	4908
3715	38097	1652	1	3854
3770	51505	1652	2	3654
5198	36219	1652	1	4700
2836	51319	1652	2	2246
	0 1474 1799 4087 3443 3656 5237 4000 4880 3829 5640 5280 4113 4866 3283 4725 3715 3770 5198	0 0 1474 19259 1799 25897 4087 51282 3443 59337 3656 69927 5237 66001 4000 46909 4880 66550 3829 50732 5640 41439 5280 53192 4113 42367 4866 42697 3283 39179 4725 53867 3715 38097 3770 51505 5198 36219	0 0 1652 1474 19259 1652 1799 25897 1652 4087 51282 1652 3443 59337 1652 3656 69927 1652 5237 66001 1652 4000 46909 1652 4880 66550 1652 3829 50732 1652 5640 41439 1652 5280 53192 1652 4113 42367 1652 4866 42697 1652 3283 39179 1652 3715 38097 1652 3770 51505 1652 5198 36219 1652	$\begin{array}{cccccccccccccccccccccccccccccccccccc$