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1 **Model selection and evaluation for risk assessment of dioxin contaminated sites**

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8 **ABSTRACT**

9 The general European population has a total intake of dioxins and dioxin-like
10 chemicals near the limit recommended by the European Union, making additional
11 exposure above background levels undesirable. For populations living near dioxin
12 contaminated sites additional exposure may occur by intake of locally produced food,
13 inhalation of particles, dermal contact with soils, or by other exposure pathways. Risk
14 assessment tools are required to estimate risks associated with contaminated sites and
15 to set priorities for site remediation. Here, we review several multimedia models that
16 can be applied as tools to support risk assessment. We then and present a strategy to
17 select, apply, evaluate and adapt a model to address a specific situation. The case
18 study we consider is a risk assessment of generic background dioxin exposure in
19 Sweden, and we compare the predictions with environmental observations and
20 exposure data from Sweden. Arguments are presented for selecting the CalTOX
21 model for this case study. We demonstrate the application, evaluation and adaptation
22 of the model, and discuss the requirements for extending the analysis to conduct risk
23 assessment for subpopulations living near dioxin contaminated sites.

24

25

26 **INTRODUCTION**

27 In the process of assessing risk and setting priorities for site remediation it is
28 beneficial to find and further develop site-specific risk assessment tools that are

1 flexible and predictive. These tools should also be able to generate information on the
2 relative significance of exposure routes. In this paper we identify and review several
3 multimedia models that can be used as decision-support tools for risk assessment at
4 contaminated sites. As part of a risk assessment project with the goal to compare
5 Swedish background dioxin exposure with exposure occurring at a dioxin
6 contaminated site, we present arguments for selecting of one of the models (CalTOX)
7 for the actual investigation. Finally, we demonstrate a case study where CalTOX was
8 applied, evaluated and adapted to describe a dioxin exposure scenario at a generic
9 Swedish background site.

10 We used the acronym PCDD/F to represent two groups of chemical
11 compounds, polychlorinated-*p*-dioxins (PCDDs) and polychlorinated dibenzofurans
12 (PCDFs). PCDD/Fs are identified as Persistent Organic Pollutants (POPs) under the
13 Stockholm Convention, and thus are internationally acknowledged to resist
14 degradation in the environment for long periods, to become widely distributed
15 geographically and to accumulate in living organisms. Among the recognized POPs,
16 PCDD/Fs are distinguished by extremely high toxicity through a specific mode of
17 toxic action-binding to the **aryl hydrocarbon or Ah receptor (1, 2). Health**
18 organizations have assigned high cancer potencies to these compounds (1). The most
19 toxic single chemical among the PCDD/Fs is 2,3,7,8-tetrachloro-*p*-dioxin (TCDD).
20 The effective toxicity of a mixture of PCDD/Fs is often expressed as TCDD toxic
21 equivalents (TEQs) (2). In this work, we used the Toxic Equivalence Factors (TEFs)
22 established by the World Health Organization (WHO) in 1998 to quantify toxicity
23 using what we refer to as WHO-TEQs (3). These TEFs were recently reevaluated and
24 revised slightly (4). However, TEQ-data that were used for comparative purposes in

Comment [TM1]: Since Ah is an abbreviation, I thin we should explain it meaning on first use.

1 this study were based on the original WHO-TEFs, and therefore we also used the
2 original factors.

3 Minimizing PCDD/F contamination of the environment has been a goal of the
4 public and governments for several decades. In the 1980s, PCDD/F emissions from
5 combustion processes and the pulp and paper industry were a major focus, and strong
6 measures were taken to reduce these primary sources. Today, secondary sources, such
7 as mobilization from previously contaminated soils and sediments are receiving
8 increasing attention. There are approximately 500 PCDD/F contaminated sites in
9 Sweden (5), most of which are former sites of wood preservation activities. The
10 chlorophenol agents that were used to protect the wood contained PCDD/F impurities.

11 Although chlorophenol agents were banned for use in Sweden in the late
12 1970s, the current content of PCDD/Fs in soils at saw-mills and wood preservation
13 sites has been estimated to be 2-50 kg TEQ (5). This can be compared to total
14 emissions of PCDD/Fs to air in Europe, which have been estimated to be
15 approximately 6 kg TEQ per year in 2000 (6). PCDD/Fs have low vapor pressure, low
16 water solubility and high affinity for particles. They are therefore not easily mobilized
17 from contaminated sites and will remain in the soil for decades or longer. While a
18 typical background soil contains a few ng TEQ/kg dry weight (dw), levels up to
19 several ten thousands ng TEQ/kg dw are not unusual at contaminated sites (7) and
20 levels as high as several hundreds of thousands ng TEQ/kg dw have been found (8).

21 The dominant exposure pathway for human exposure to PCDD/Fs is intake via
22 food. The European Commission estimates that food is responsible for at least 95% of
23 total intake (9). It has been shown that the exposure of humans may be of concern
24 even at background levels (10). The tolerable weekly intake (TWI) of PCDD/Fs and
25 chemicals with dioxin-like toxicity (dioxin-like PCBs) is 14 pg TEQ per kg

1 bodyweight (bw) as recommended by the European Union (11), and this
2 recommendation is commonly referred to as a tolerable daily intake (TDI) of 2 pg
3 TEQ per kg bw. The average intake of PCDD/Fs and dioxin-like PCBs by adults has
4 been estimated to be 1.2-3.0 pg TEQ/kg bw/day for European countries, and the
5 average daily intake by the Swedish population was estimated to be in the same range
6 (average 1.4 pg TEQ per kg bw; 95th percentile 3.0 pg TEQ per kg bw) (12). Thus,
7 even when considering only background PCDD/F contamination, a considerable
8 proportion of the European and Swedish populations have a daily intake close to or
9 above the recommended limit.

10 Based on the information above, it is apparent that additional PCDD/F
11 exposure above background levels can pose a health risk. For populations living near
12 contaminated sites, additional exposure may occur by intake of locally produced food,
13 inhalation of particles, dermal contact with soils, or by other exposure pathways.
14 Remediation of all of the chlorophenol contaminated sites in Sweden would eliminate
15 these exposures; however the estimated costs significantly exceed available resources.
16 Therefore, informative risk-benefit analyses are needed to prioritize actions for
17 mitigating the elevated risk at specific sites through actions to remediate the site or
18 block exposure pathways. In this process, models can provide a quantitative
19 framework to support decision making. The goals of this paper are to identify and
20 select an appropriate model and to demonstrate how model evaluation can be used to
21 build confidence that the selected model is a useful tool to support decision making.

22

23 **MODELING TOOLS**

24 We conducted search to identify frequently used and easily available models
25 appropriate for regional- or local-scale risk assessment that could be applied to

1 contaminated sites. This search identified five models that have been applied in
2 Europe in the past. In this section, we provide a brief summary of these models and
3 compare and contrast their features. All models discussed here are multimedia box
4 models, which calculate the distribution of a chemical among various environmental
5 compartments on the basis of chemical partitioning (13). To carry out these
6 calculations, physicochemical properties of the compound and properties of the
7 environment are used. Another common feature of the models is that human exposure
8 is modeled by considering a number of potential exposure pathways. A comparison of
9 the direct and indirect pathways included in each model is presented in Table 1. Direct
10 exposure pathways describe exposures to the substance in a medium to which it was
11 first released. Indirect exposure pathways are those for which there is at least one
12 intermedia transfer, or intermediate biological transfer step between the source and
13 the points of exposure. Below is a short review of the five models. Each has its
14 advantages and disadvantages, and the choice of model depends on the aim of the
15 user.

16 *The Swedish Model (NV):* The Swedish model for general and site specific
17 guideline values for soils was developed at the request of the Swedish Environmental
18 Protection Agency (14). The calculations are based on soil contamination and
19 partitioning of the chemical between the soil, pore-water and pore-air. From these
20 compartments, the contaminants can be transported to other compartments
21 (groundwater, surface water and air), and concentrations in receiving compartments
22 are calculated. Human exposure is then calculated from these environmental
23 concentrations and dilution factors via selected exposure routes (Figure 1, Table 1).
24 The model is static, and does not include prediction of future changes, such as
25 declining levels due to degradation and transport out of the model world.

1 *CLEA*: The Contaminated Land Exposure Assessment Model (CLEA) was
2 developed for the British Environmental Agency (15). It was created for the purpose
3 of estimating soil guideline values and site specific assessment criteria. The model
4 world consists of soil and air, and the only contamination source is secondary
5 emissions from soil. It is a static model since contaminant degradation over time is not
6 considered. There are 10 exposure routes (Table 1) and the calculated exposure is
7 compared to certain health criteria such as tolerable daily soil intake. In the case of
8 site-specific assessment, recommendations are given on whether further actions are
9 needed.

10 *CSOIL*: This model was developed by the National Institute for Health and the
11 Environment for the Ministry of the Environment, the Netherlands (16-18). It was
12 designed to assess soil and ground-water quality and is a progenitor of the Swedish
13 model. It calculates the distribution of contaminant between soil, pore-water and pore-
14 air. Contaminants may be transported to groundwater, outdoor air and indoor air.
15 Similar to the Swedish model and CLEA, the model is designed to handle a pool of
16 contaminants present in the soil, but no other primary or secondary sources are
17 considered, and neither are loss processes; it is thus also a static model. CLEA
18 accounts for potential human exposure through seven different pathways (Table 1). It
19 can be applied in a site-specific way to determine remediation urgency.

20 *EUSES*: The European System for the Evaluation of Substances (EUSES) has
21 been developed within the European Union for risk assessment of new and existing
22 substances and biocides (19). This model is intended to be used as a screening tool at
23 a personal scale (for consumers and workers), at the local scale (for human
24 populations and ecosystems near point sources) and at regional scale (for human
25 populations and ecosystems exposed as a result of releases in a region). It has two

1 modules; one that estimates the distribution and transport of the chemical in
2 environmental compartments (Predicted Environmental Concentrations; PECs), and
3 one that calculates the human and environmental exposures from this scenario.
4 EUSES includes air, surface water, ground water, soils and sediment compartments. It
5 can accommodate continuous primary as well as secondary source emissions.
6 However, the model is not designed for initial soil concentrations. EUSES is a steady
7 state model (mass fluxes are constant over time) that considers contaminant
8 degradation and other outflows. Most of the human exposure pathways that are
9 considered are indirect pathways via food (Table 1).

10 *CalTOX*: CalTOX is a multimedia mass balance model that was developed
11 originally by McKone (20) for the California Environmental Protection Agency to
12 assess risks in connection with hazardous waste sites. Similar to EUSES, it integrates
13 a local/regional multicompartment fate model with a multi-pathway human exposure
14 model. The user can specify either continuous emissions to air, soil or surface water,
15 or a one-time initial concentration in soil representing in-place contamination. Unlike
16 the other models reviewed here, CalTOX can be used for calculating human exposure
17 from contaminated soil in which the concentrations decline over time. It is, however,
18 not a fully dynamic model since emission incidents can not be introduced other than
19 at the initial stage. The exposure pathways include most conceivable exposure routes
20 (Table 1).

21

22 **THE PCDD/F BACKGROUND SCENARIO FOR SWEDEN: AN EXAMPLE** 23 **OF MODEL SELECTION, APPLICATION AND EVALUATION**

24 In this section we describe the selection of the most suitable model, and the
25 requirements for building confidence in the model. Using our selected model, we

1 demonstrate a process of model evaluation and adaptation to achieve satisfactory
2 model performance for the scenario of human exposure at a Swedish background site.

3

4 **MODEL SELECTION**

5 The first task in our investigation was to select the most appropriate model for
6 risk assessment of PCDD/F exposure for populations living at or near contaminated
7 land as compared to populations living at background sites. We carried out this
8 process with the recognition that all models have inherent capabilities and limitations.
9 But it was also important that the model selected for the PCDD/F risk assessment
10 contain a minimum set of capabilities that we identified as necessary to capture both
11 the potential magnitude and variation of exposures near contaminated lands as well as
12 background exposure. Based on the aims of our risk assessment project, it was
13 desirable that the selected model should support:

- 14 i) continuous emissions to air and water
- 15 ii) initial concentrations in soil and sediment
- 16 iii) estimation of changes over time (dynamic model)
- 17 iv) estimation of human exposure via all pathways that are potentially
18 relevant for PCDD/Fs in situations with elevated soil concentrations,
19 i.e. via direct as well as indirect exposure routes.

20 Even though CSOIL, CLEA and the Swedish model were all designed to
21 support decisions in risk assessments of contaminated land, none of these models
22 were designed to handle a complete range of multiple transport and/or exposure
23 pathways as a consequence of a given source strength in the environment. Among
24 EUSES and CalTOX, only CalTOX provided a possibility to treat both emissions to
25 air and water as well as an initial soil concentration. Another factor that favored the

1 use of CalTOX was its ability to fully address long-term mass balance gains and
2 losses by chemical degradation processes in each compartment and by transport to
3 and removal with soil erosion, surface water outflow, and ground water movement.
4 Although CalTOX is not a truly dynamic model, it takes degradation into
5 consideration and initial concentrations in soil are depleted over time due to transport
6 and transformation processes.

7 The models also differed in the exposure pathways that are considered (Table
8 1). EUSES does not focus on direct exposure pathways, which are of importance
9 when considering exposure from contaminated soil, and the NV, CSOIL and CLEA
10 models do not consider some food chain pathways that are of importance for
11 estimating PCDD/F exposure (i.e. intake of eggs and cow's milk). In summary,
12 CalTOX was the model that offered the best fit to the requirements that were set for
13 our investigation.

14 However, CalTOX also has limitations. Although it allows dynamic modeling
15 in soil layers, it does not allow for dynamic mass balance in air, surface water, and
16 sediments and it does not allow for dynamic incidents other than initial loadings and
17 continuous emissions. Another limitation in CalTOX (as well as the other models) is
18 the inability to introduce an initial pool of contaminants in sediments. Such a
19 secondary source is likely to be significant in many real-world PCDD/F
20 contamination situations where land areas for former industrial activities are in close
21 proximity to surface waters.

22 CalTOX has been examined by experts in the field, and it has been applied in
23 several scientific studies (21-26). The design of the model is shown in Figure 2 and a
24 compilation of the processes that are considered is shown in Table 2. The model can
25 be used for any substance for which partition coefficients, biotransfer factors,

1 bioavailability, and degradability are known. Deterministic calculations can be carried
2 out using single-value inputs producing single-value exposure estimates, and
3 probabilistic results (mean and variance of exposures) can be obtained when one or
4 more of the model inputs are specified using a range of values selected from a
5 distribution of input values. The model is structured to carry out parameter sensitivity
6 analysis and uncertainty importance ranking. CalTOX includes support for Monte
7 Carlo uncertainty analysis using the software Crystal Ball[®] (Decisioneering, Inc.,
8 Denver, CO, USA), but there are several options for conducting sensitivity and
9 uncertainty analyses.

10 There are three categories of input data that are needed for CalTOX: physico-
11 chemical data ($n=35$), landscape parameters ($n=58$) and population exposure-factor
12 parameters ($n=54$). Since CalTOX has a default option of estimating many parameter
13 values from other parameters, not all of the input data has to be specified by the user.
14 For example, many of the bioconcentration factors in CalTOX can be estimated using
15 the octanol-water partition coefficient; hydrologic parameters, such as evapo-
16 transpiration and runoff, can be estimated from rainfall; and many of the human-
17 activity parameters can be set at default values that are representative of norms in
18 industrialized countries.

19

20 **BUILDING CONFIDENCE IN A MODEL AND ITS RESULTS**

21 Identifying a model that meets the user's selection criteria is only the first step
22 in the process, and is not adequate to assure that the model will provide useful
23 information for risk assessment decisions. Any model, even one that is widely used
24 and commonly cited, should be subjected to a performance evaluation that builds
25 confidence in the appropriateness of the model to support a specific decision. For

1 example, Bonnard (26) reviewed the use of CalTOX in French risk assessment
2 investigations, and showed that in spite of the many advantages of the model and its
3 value for assessing waste sites in France, there are many common errors that have
4 been made in applying this model. These errors cause loss of confidence in risk
5 assessments based on models. Bonnard suggests that many of the errors are the result
6 of using the model as a “black box”, and that this problem can be avoided with a
7 model evaluation process.

8 Here, we divide model evaluation into three steps: 1) obtaining appropriate
9 input data, 2) evaluation of model output values by comparing with measurement
10 data, and 3) adapting the model and re-evaluating input data to achieve better
11 agreement with real world observations. Deviations between the model and
12 observations can be due to applying an inappropriate model or lack of reliability in
13 measurements and input data. In the sections below, we consider some aspects of the
14 performance evaluation that can increase or decrease the confidence about the results
15 obtained from a model.

16 *Selecting input data.* Reliable model performance can only be expected if
17 input data are accurate. In particular, selecting physicochemical properties values to
18 describe the chemical may be challenging since there is generally a range of measured
19 and calculated values in the literature (27). In many modeling studies, property values
20 are selected on the basis of expert recommendations. But to increase confidence in the
21 model output values, it is advisable to select chemical properties data using methods
22 that exploit the relationships between partition coefficients to identify values that may
23 be in error, and in some cases to calculate a single set of values that are internally
24 consistent (28, 29). For our case study we reviewed the literature and evaluated the
25 internal consistency of our selected partitioning properties. However, there are large

1 uncertainties about the true values of the partitioning properties of the PCDD/Fs due
2 to wide variability in reported values and a degree of expert judgment is required to
3 select appropriate values.

4 *Comparison with measurement data.* Environmental fate models are
5 incomplete descriptions of open systems consisting of the chemical and the
6 environment (30). Therefore they cannot be validated using methods that work for
7 models that describe completely closed, controlled systems. However, in any risk
8 assessment it is important to evaluate the performance of the model by comparing
9 predicted environmental concentrations and exposures with available data. Ideally, a
10 model evaluation and adaptation process is conducted for a site specific scenario for
11 which recent environmental measurements have been conducted. However, even if
12 the temporal and spatial scale is not always a perfect match, this kind of evaluation is
13 necessary and can help to find weaknesses. It is also advantageous for the model
14 evaluation to include chemicals with a range of physico-chemical properties.

15 *Adapting the model and re-evaluating input data to address differences*
16 *between model output and measurements.* An important step in the model evaluation
17 process is to scrutinize the model output values that do not agree with relevant
18 measurement data that is believed to be reliable. Quality of input data as well as
19 usefulness of algorithms must be evaluated. In this process, it is advantageous to start
20 with sensitivity and uncertainty analyses. These procedures help to identify
21 parameters with a large impact on the results. A complete description of these
22 methods is beyond the scope of this summary paper, but the widely cited text by
23 Morgan and Henrion (31) can be consulted for a summary and suggestions. Once the
24 critical parameters and model algorithms are identified, efforts can be focused on
25 refining the input data and model algorithms as much as possible.

1

2 **MODEL APPLICATION, EVALUATION AND ADAPTATION**

3 After the model selection, we created a background scenario for Swedish
4 conditions and modeled human exposure by using the selected model (CalTOX). This
5 scenario is intended to be the base case, which can be compared to a similar scenario
6 in which a highly contaminated soil is introduced. Our case study considered six tetra-
7 through octa-chlorinated dioxin and furan congeners (2,3,7,8-TCDD, 2,3,4,7,8-
8 PeCDF, 1,2,3,6,7,8-HxCDD, 1,2,3,4,6,7,8-HpCDD, 1,2,3,4,6,7,8-HpCDF and
9 OCDD). The congeners were selected to represent i) a broad range of
10 physicochemical properties of PCDD/Fs (2,3,7,8-TCDD and OCDD), ii) congeners
11 that commonly contribute significantly to the TEQ-value at contaminated sites
12 (1,2,3,6,7,8-HxCDD, 1,2,3,4,6,7,8-HpCDD and 1,2,3,4,6,7,8-HpCDF), and iii) a
13 congener that is relatively abundant in food (2,3,4,7,8-PeCDF).

14 We modeled a land area of 10,000 m², and assumed that there was surface
15 water with an average depth of 3 m covering 1000 m² of this area. The modeled area
16 reflects the typical size of a contaminated site in Sweden. The mixing height of the
17 atmosphere over the modeled area is assumed to be 9 m and the residence time of air
18 is 100 seconds. Climate data was selected to represent southern Sweden and other
19 landscape parameters were matched as much as possible to available data from
20 Swedish and Norwegian investigations. The modeled soil system consists of three
21 layers (from top to bottom); the surface layer (2 cm), the root layer (1 m) and the
22 vadoze zone (1 m), and the organic carbon content for each layer (from top to bottom)
23 were set to 3.6%, 3.6% and 0.4 %, respectively. The thickness of the sediment layer
24 was set to 5 cm and its organic carbon content to 2.9%.

1 It is generally believed that the most important current PCDD/F releases are
2 air emissions from current activities and/or reservoir sources (32). As a result, our
3 background scenario was based on continuous air emission and an in-place reservoir
4 in soils as PCDD/F sources. The emission parameters were adjusted to levels at which
5 the calculated air and surface soil concentrations were in good agreement with
6 average concentrations measured at background sites in southern Sweden since the
7 year 2000 (33, 34).

8 Measurements of PCDD/Fs are expensive and often excluded from national
9 environmental monitoring programs. For evaluating our background scenario, we
10 were obliged to resort to relatively scattered data from the peer-reviewed literature,
11 Swedish reports and also some unpublished (in-house) values. As described below,
12 Figures 3 and 4 show monitoring data for PCDD/Fs from background sites in and
13 around Sweden that we used to evaluate the model results. The modeled data in
14 Figures 3 and 4 is the final result obtained after model adaptation and re-evaluation of
15 input data. However, in our initial model evaluation based on our first selected input
16 data and the default model algorithms, the model underestimated PCDD/F
17 concentrations in vegetation by approximately an order of magnitude relative to
18 measurement data, and aquatic and the terrestrial food chain transfers of the PCDD/Fs
19 were found to be out of agreement with measurement data by up to several orders of
20 magnitude. The unsatisfactory agreement between modeled and measured
21 environmental concentrations and exposures was evidence of inappropriate input data
22 and model limitations that required actions to improve model performance. Here, we
23 summarize our experience, which provides an example of how to overcome initial
24 prediction errors.

1 *Concentrations of PCDD/Fs in vegetation.* With its default parameterization,
2 the CalTOX model systematically underestimated concentrations of PCDD/Fs in
3 vegetation leaves as compared to available data for grass gathered at a background
4 site in Sweden. Analysis of the model results indicated that the calculated
5 concentration in whole leaves was being controlled by the assumed rate of cuticle
6 erosion, which transfers PCDD/F to the soil from the plant surface. The half-time for
7 this process used by default in the model is 14 days (35), which is an estimate based
8 on a consensus opinion of experts. However, more recent experiments that studied
9 the uptake of semi-volatile organic chemicals by grass from air indicate that at least
10 six weeks is required to approach steady-state conditions (36). We therefore changed
11 the model input value for half-time for cuticle erosion to 42 days, which dramatically
12 improved agreement between the model and measurement data.

13 *The terrestrial food chain transfer of PCDD/Fs.* Critical parameters for meat,
14 cow's milk and egg concentrations are the various biotransfer factors (BTFs), which
15 describe the degree of uptake of the contaminant in animals and the rate of transfer
16 from feed to animal and further to food products (egg, milk). At an early stage of the
17 modeling, we noted that the CalTOX model overestimated the meat and cow's milk
18 exposure routes for highly chlorinated PCDD/Fs. The default algorithms for these
19 pathways are based on work by Travis and Arms (37), which assume linear
20 relationships between the logarithm of biotransfer factors and the logarithm of the
21 octanol-water partition coefficients (K_{ow}). However, the bioavailability for super
22 hydrophobic compounds ($\log K_{ow} > 6$) is substantially limited. Thus the highly
23 chlorinated PCDD/Fs are less available for uptake, and the BTFs decline with
24 increasing hydrophobicity for highly chlorinated congeners. For meat and cow's milk,
25 a better agreement between predicted and measured values was achieved by using

1 experimentally determined BTFs (38) rather than default algorithms in CalTOX. For
2 eggs, no experimental BTFs could be found in the literature, and we were therefore
3 relied on the default algorithms. The congener patterns and absolute levels of the
4 modeled meat and milk exposures were in good agreement with those found in the
5 food survey, in both cases within a factor of 4 of reported exposures for 7 out of 12
6 comparisons and in all cases within a factor of 8, except for OCDD in meat
7 (prediction 24 times below observed value; Figure 4). Egg exposure was clearly
8 under-predicted for the lower chlorinated congeners, but agreement for the higher
9 chlorinated congeners was within a factor of 3.

10 *The aquatic food chain transfer of PCDD/Fs.* The limited bioavailability of
11 super hydrophobic chemicals is also not considered in the default description of the
12 bioconcentration process from water to fish in CalTOX. Similar to the terrestrial food
13 chain transfer, a better agreement between predicted and measured values was
14 achieved if the CalTOX default BCFs were replaced by measured bioconcentration
15 factors (BCFs), and the best fit was obtained by using BCFs published by Govers and
16 Krop (39) (Figure 4). Analysis of fish from Swedish lakes with background levels
17 generally showed PCDD/F concentrations below the limit of detection, and therefore
18 a comparison was made with fish from a lake that is located in an urbanized region in
19 Sweden. The average value of six composite samples of salmon (*Salmo salar*), trout
20 (*Salmo trutta*) and char (*Salvelinus alpinus*) from Lake Vättern was used (two
21 samples per species; 7-10 specimens in each composite) (40). The individual fish
22 weights ranged from 0.5 g to 3.7 kg (average 1.9 kg) and the lipid weights from 1.3 to
23 4.6% (average 2.9%). The observed values were significantly higher than the
24 predicted (Figure 4). The large deviation between observed and predicted values is
25 partly attributable to the difficulties in finding relevant comparison data, but primarily

1 to the use of the water-concentration BCF product in CalTOX to calculate
2 contaminant concentrations in fish. This approach neglects the bioaccumulation of
3 contaminants by high-trophic-level fish via intake of lower trophic level prey. For
4 high trophic level fish it is more appropriate to use bioaccumulation factors (BAFs).
5 The use of BAFs observed for adult lake trout (*Salvelinus namaycush*) from Lake
6 Ontario (41) instead of using BCFs resulted in good agreement between predicted and
7 measured values. The lipid fraction of the Lake Ontario trout was 8%, and therefore
8 the BAF-values shown in Figure 4 were down-adjusted to match a lipid fraction of
9 3% in order to better compare with the Lake Vättern data.

10 The predicted values in the CalTOX model are on wet weight basis, and there
11 is no option in the model to distinguish between lean and fat fish. In light of all these
12 issues, it is clear that modeling “average” fish concentrations is a challenge. A step
13 towards better predictions is to expand the fish model so that differentiated
14 information on the fish that is consumed can be used. From the fish concentration
15 modeling perspective, it would also be beneficial to adjust the model so that it allows
16 initial dioxin concentration in the sediment. The sediments often act as significant
17 secondary sources to the water column and to the benthic food chain due to
18 historically higher emissions, thereby also influencing the fish levels and the exposure
19 from fish ingestion.

20 As stated earlier, a more complete and comprehensive model evaluation would
21 be possible with consistent site-specific measurements and modeling. However the
22 findings described above are an example of the benefit of conducting model
23 evaluation even for generic scenarios. Our experience also highlights the advantage
24 of including congeners with a range of physico-chemical properties for the evaluation
25 of a modeled scenario.

1

2 **PERFORMANCE OF THE ADJUSTED MODEL**

3 In Figure 3, we compare the final, adapted model output for concentrations in
4 vegetation, surface water, and wet and dry atmospheric deposition to Swedish
5 reference data for three of the congeners (2,3,7,8-TCDD, 1,2,3,6,7,8-HxCDD,
6 1,2,3,4,6,7,8-HpCDF). Figure 3 also shows the background concentrations in air and
7 surface soil that were fitted as emission parameters. The model output and observed
8 concentrations/rates in most cases agreed within a factor of 4. The exception is the
9 atmospheric deposition rate of 2,3,7,8-TCDD, which the model underestimates by an
10 order of magnitude. The extent of gas-particle partitioning of 2,3,7,8-TCDD is
11 sensitive to partitioning properties, and the observed discrepancy could be attributable
12 to errors in the specific values selected for the study. However, considered as a
13 whole, Figure 3 indicates that the model results are in very good agreement with
14 observations, especially considering that the monitoring data has been collected from
15 various sites around Sweden.

16 The corresponding exposure modeling was based on average food
17 consumption values compiled by the National Food Administration in Sweden (12,
18 42). We used the default intake rates for an adult population and modeled the PCDD/F
19 exposure near the background site described above. The predicted TEQ-values of
20 direct and indirect exposures from the adjusted model are shown in Table 3, and
21 exposures via specific food pathways are shown in Figure 4. Predicted direct exposure
22 contributed only 1.5% of the total. This prediction is near the estimations from
23 European measurement data, which indicate that direct exposure generally accounts
24 for only 2-5% of the total exposure (9).

1 The indirect PCDD/F exposure (intake via food, including PCDD/Fs and no
2 other dioxin-like compounds) for the Swedish population has been estimated to be
3 0.05-12 pg TEQ/kg bw per day (min and max values) and on average 0.8 pg TEQ/kg
4 bw per day by analyzing a typical Swedish “food basket” (12). The comparison
5 between the modeled food exposures in Table 3 and these food basket values is
6 problematic since in the food basket study, all of the seventeen 2,3,7,8-chlorinated
7 congeners were included, and in the modeled TEQ-value only six of these highly
8 toxic congeners were considered. We estimated that these limitations will lead to an
9 underestimation of TEQ intake by a factor of 2. Thus, in a more complete comparison
10 with the national food intake value, the modeled value should be doubled. By
11 considering this, the predicted value (0.12 pg TEQ/kg bw per day) is within the range
12 estimated in the food basket survey, and 7 times lower than the Swedish average value
13 (Table 3). We view this as reasonable given that the model scenario represents a
14 background site and does not include exposure from consumption of fish with
15 elevated concentrations, which is known to be an important exposure pathway for the
16 general population of Sweden (12). It should also be noted that food-basket surveys
17 also are subject to uncertainties and variability in their estimates of population
18 exposure.

19

20 **DISCUSSION AND CONCLUSIONS**

21 The case study described here clearly indicates that model selection is only the
22 beginning of a successful and informative modeling process. Once a model is
23 selected, a rigorous process of model evaluation is needed to build confidence in the
24 performance of the model in the context of the specific issue under consideration. For
25 the case study, satisfactory agreement between modeled values and available

1 measurement data was demonstrated after the model evaluation and adjustment
2 process. However, it also became clear that site specific measurements and renewed
3 evaluations and adjustments are required in order to establish sufficient confidence in
4 the model to function as a reliable risk assessment tool for human PCDD/F exposure
5 near contaminated sites.

6 The case study presented here supports our selection of the CalTOX model as a
7 tool that can quantitatively inform risk assessment when applied in conjunction with
8 appropriate monitoring data that enables model evaluation. Our experience provides
9 several instructive lessons about the capabilities and limitations of a model-based site-
10 specific risk assessment. In particular, risk assessors must be aware that confidence in
11 model results can only be built based on successful model evaluation for cases similar
12 to the one of interest. This requires a foundation of reliable input data for chemical
13 properties and landscape parameters, and comparison of the model results with
14 monitoring data appropriate to the space and time scales of the assessment. In our case
15 study, we found that property data, landscape parameters and the default model
16 algorithms can all introduce significant uncertainties in the assessment. To achieve
17 satisfactory agreement between the model and monitoring data, careful selection of
18 input data and adjustments to the model algorithms were required, namely using
19 updated physiological parameters for cuticle erosion, using experimentally determined
20 bioconcentration factors or BAFs derived from field studies rather than default
21 algorithms in CalTOX, and using experimentally determined biotransfer factors rather
22 than default algorithms for calculation of PCDD/F concentrations in cow's milk and
23 beef.

24 This work shows that models should not be used blindly as a "black-box"
25 source of information in risk assessments. There is a strong incentive to develop and

1 implement models with high predictive power, especially since quantitative risk
2 assessment facilitates the formulation of risk management strategies. However,
3 applying and interpreting chemical fate and exposure models requires expertise and
4 experience. Decision makers should not expect models to provide unequivocal
5 answers, and sound management strategies must be based on a combination of
6 quantitative modeling and expert judgment. In the case study presented here, we
7 provide an example of a sequential process that includes model selection, input data
8 quality evaluation, collection and interpretation of observations, model evaluation,
9 and model adaptation. Our results show that this process offers insight and knowledge
10 that neither a model nor environmental data alone can provide.

1 Table 1. Exposure pathways included in selected risk assessment modeling tools
 2 (modified from Rikken *et al.* (17); consumption of egg and supplementary
 3 details for Swedish model (NV) were added).
 4

		EUSES	NV	CSOIL	CLEA	CalTOX
Direct exposure	Ingestion of contaminated soil particles (outdoor)		x	x	x	x
	Ingestion of contaminated soil particles / dust (indoor)		x	x	x	
	Dermal contact with soil contaminants (outdoor)		x	x	x	x
	Dermal contact with soil contaminants / dust (indoor)		x	x	x	
	Inhalation of contaminated soil particles / dust (outdoor)		x	x	x	x
	Inhalation of contaminated soil particles / dust (indoor)		x	x	x	x
	Inhalation of vapors from contaminants (outdoor)	x	x	x	x	x
	Inhalation of vapors from contaminants (indoor)		x	x	x	x
	Indirect exposure	Consumption of crops	x	x	x	x
Consumption of soil attached to crops					x	x
Consumption of fish		x	x			x
Consumption of meat		x				x
Consumption of egg						x
Consumption of milk		x				x
Consumption of contaminants via mothers' milk						x
Consumption of contaminants via drinking water		x	x	x		x
Ingestion of contaminants via bathing water						x
Inhalation of vapors from drinking water from shower					x	x
Dermal contact with drinking water from shower/bath					x	x
Dermal contact with contaminants via bathing water						x

5
 6 x: Exposure route available; x⁺ Exposure route available, but not as a separate
 7 pathway; **EUSES**: European System for the Evaluation of Substances The
 8 Contaminated Land Exposure Assessment Model (19); **NV**: The Swedish model (14);
 9 **CSOIL**: Model developed by the National Institute for Public Health and the
 10 Environment, the Netherlands (16-18); **CLEA**: Contaminated Land Exposure
 11 Assessment Model, U.K. (15); **CalTOX**: Model developed for the California
 12 Environmental Protection Agency (20)

- 1 Table 2. Summary of the processes by which contaminants are exchanged and lost
 2 among seven CalTOX compartments (20).

Compartment	Gains	Losses
Air (both the gas phase and particles of the troposphere)	diffusion from soil diffusion from plants diffusion from surface water resuspension of deposited soil particles <i>contaminant sources</i>	diffusion to ground-surface soil diffusion to surface water diffusion to plants washout by rainfall convection losses deposition to soil deposition to plants deposition to surface water chemical/physical transformation
Plants (land vegetation)	deposition of particles from air foliar uptake root-uptake from root-zone soil	diffusion from leaf surfaces wash off from leaf surfaces chemical/physical transformation
Ground-surface soil	diffusion from air diffusion from root-zone soil washout from air by rainfall dry deposition of air particles <i>contaminant sources</i>	diffusion to air diffusion to root-zone soil advection to root-zone soil soil solution runoff erosion (mineral runoff) to surface water resuspension of soil particles chemical/physical transformation
Root-zone soil	diffusion from ground-surface soil advection from ground-surface soil <i>contaminant sources</i>	diffusion to ground-surface soil infiltration (leaching) to vadose zone soil chemical/physical transformation
Vadose-zone soil	infiltration from root-zone soil <i>contaminant sources</i>	infiltration to ground-water zone
Surface water	diffusion from air washout by rainfall deposition of atmospheric particles soil solution runoff erosion (mineral runoff) diffusion from sediment sediment resuspension <i>contaminant sources</i>	sediment deposition diffusion of vapors to air diffusion to sediment surface-water outflow chemical/physical transformation
Sediment layer	diffusion from surface water sediment deposition (from surface water)	diffusion to surface water sediment resuspension chemical/physical transformation

1 Table 3. Modeled and measured PCDD/F exposure for an adult population living at a
 2 background level site (pg WHO-TEQ/kg bodyweight and day).

	Direct exposure	Indirect exposure	Total exposure
Modeled value including six of the seventeen 2,3,7,8-chlorine substituted PCDD/F congeners ^a	0.0009	0.06	0.06
Estimated value including all of the seventeen 2,3,7,8-chlorine substituted PCDD/F congeners ^b	0.002	0.12	0.12
Measured values, typical Swedish "food basket" (12)			
- range (min-max)	NA	0.05-12	NA
- average		0.8	

3 ^aCongeners included: 2,3,7,8-TCDD, 2,3,4,7,8-PeCDF, 1,2,3,6,7,8-HxCDD,
 4 1,2,3,4,6,7,8-HpCDD, 1,2,3,4,6,7,8-HpCDF and OCDD; ^bThe value was estimated by
 5 multiplying the modeled TEQ-value for six congeners by a factor of 2; NA: not
 6 analyzed

1

2 **Figure Legends**

3

4 **Figure 1.** An overview of the structure of the Swedish EPA calculation model for site
5 specific limit values for use of land.

6

7 **Figure 2.** The framework of the CalTOX model.

8

9 **Figure 3.** Modeled concentrations of A) 2,3,7,8-TCDD, B) 1,2,3,6,7,8-HxCDD and
10 C) 1,2,3,4,6,7,8-HpCDF in background scenario as compared to measured
11 concentrations of Swedish environmental compartments; air (ng/m³) (33) ground
12 surface soil (ng/kg wet weight) (34); total leaf (ng/kg dw) and surface water (ng/l;
13 unpublished data, Umeå University, Sweden) and atmospheric wet and dry deposition
14 (ng/day) (33).

15

16 **Figure 4.** Modeled and measured concentrations (mean and one standard deviation)
17 of meat, diary products/cow's milk, egg (12) and fish (Lake Vättern; salmon, trout
18 and char from the northern part; average of six pooled samples, two from each
19 species) (40) in ng WHO-TEQ/kg wet weight. BTF: Biotransfer factor; BCF:
20 Bioconcentration factor; BAF: Bioaccumulation factor. The modeled values were
21 obtained by using BTFs from Birak *et al.* (38), BCFs from Govers and Krop (39) and
22 BAFs from Morrison *et al.* (41). The modeled BAF-values were obtained by
23 assuming a lipid weight of 3%.

Figure 1

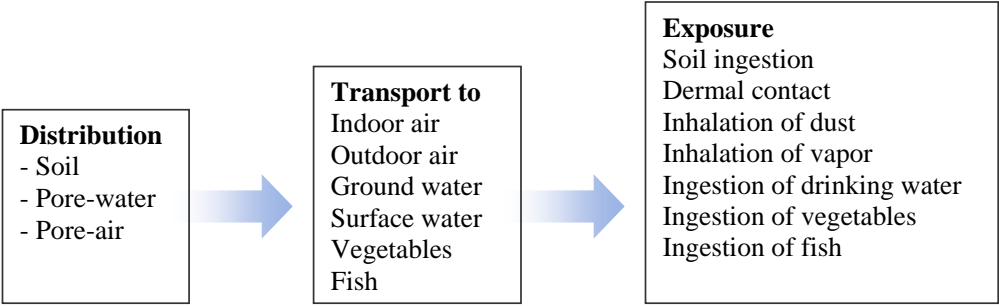
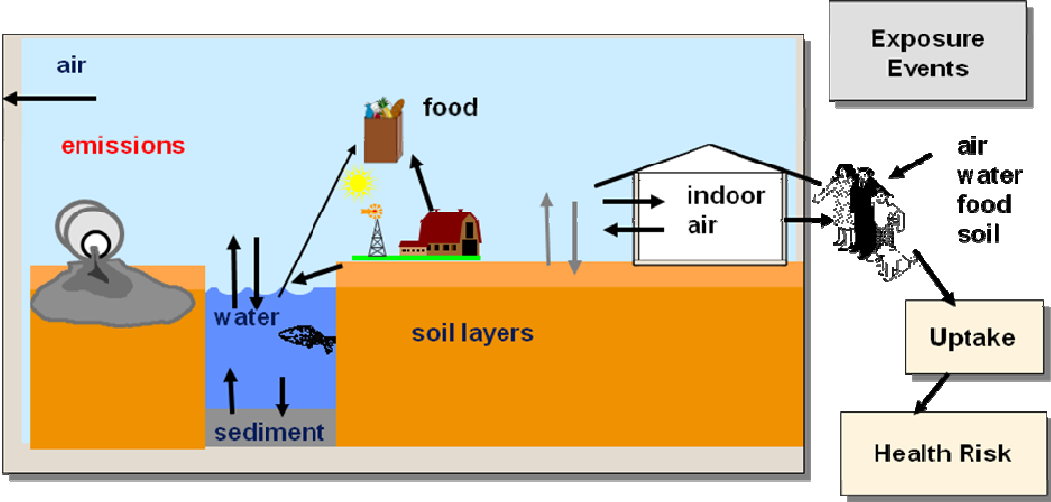
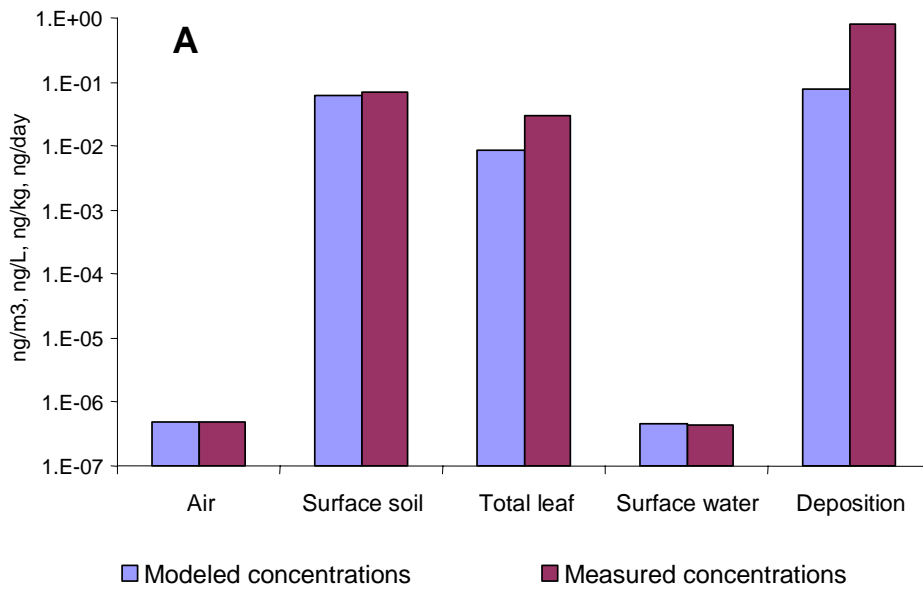


Figure 2

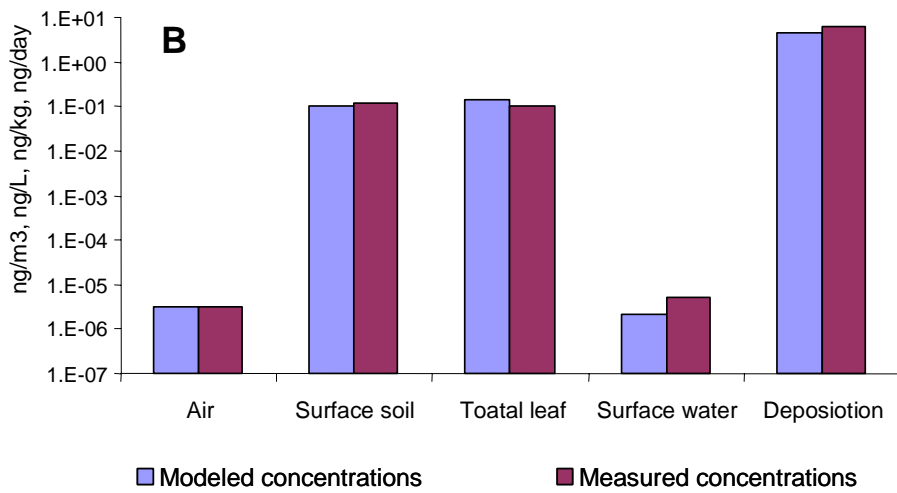


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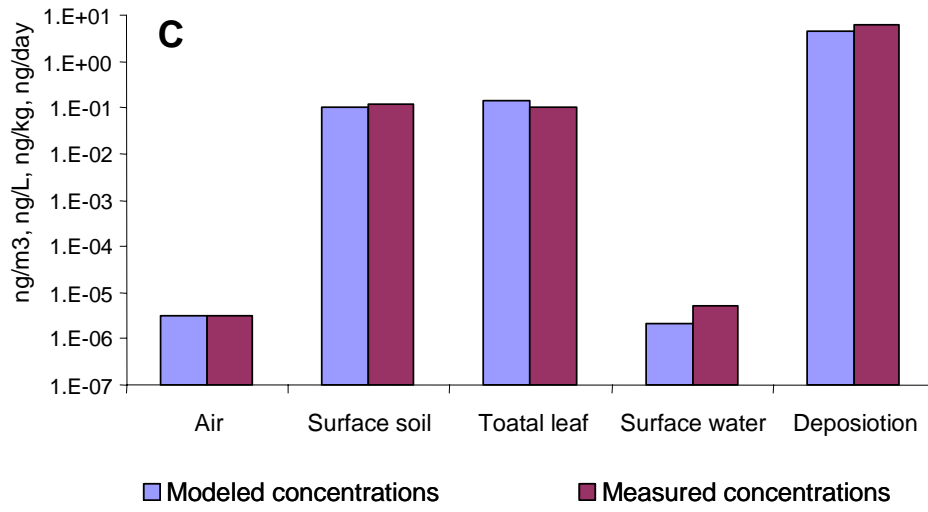
1 Figure 3



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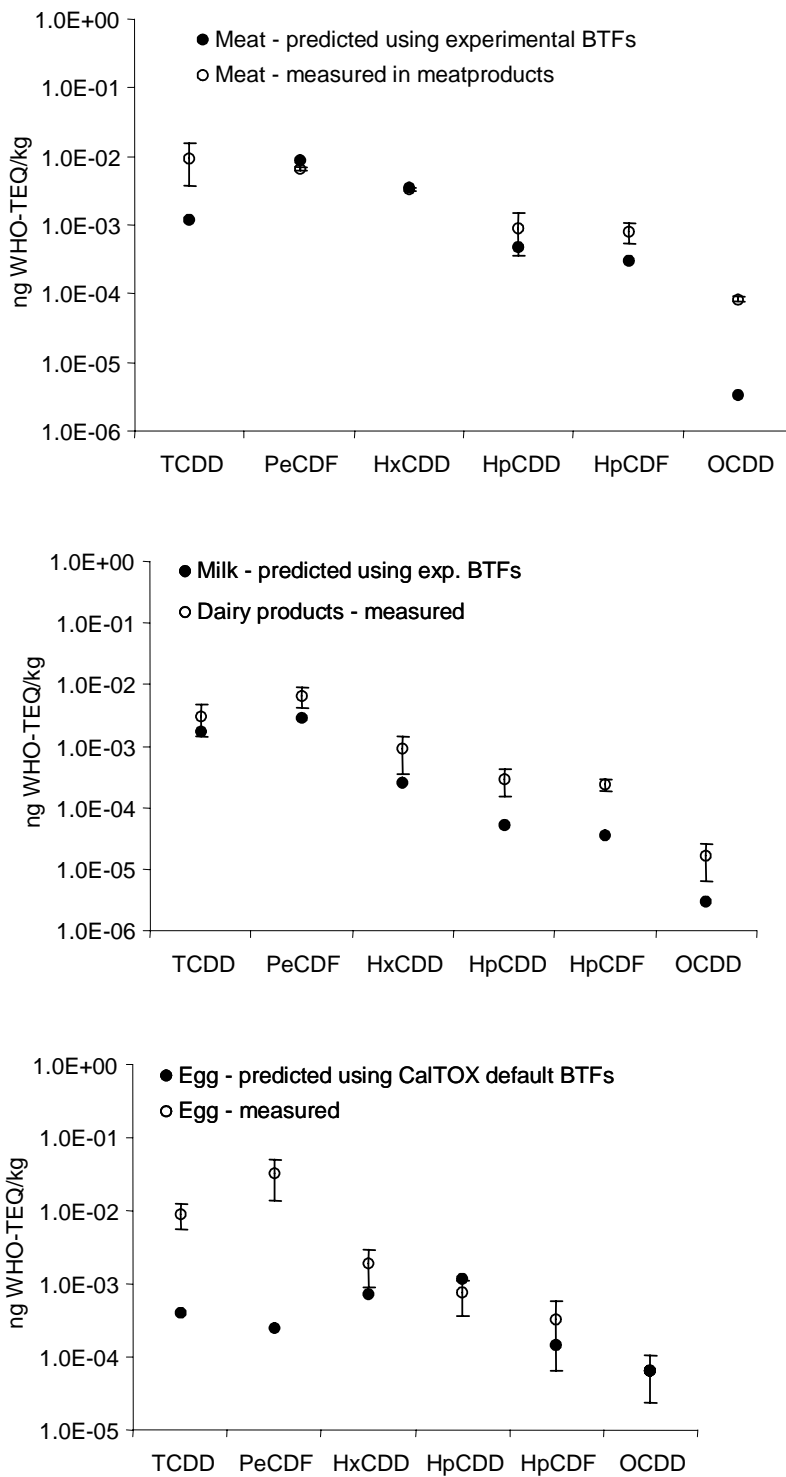


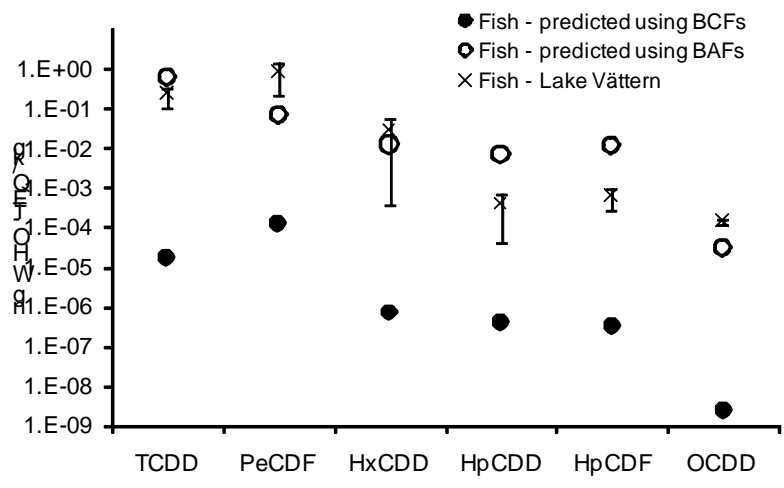
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1 Figure 4





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