



OpenFoodTox and other Open Source *In silico* Tools @ EFSA

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Scientific Committee and Emerging Risks Unit
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Summer School
"In silico methods in food safety"
14th June 2017

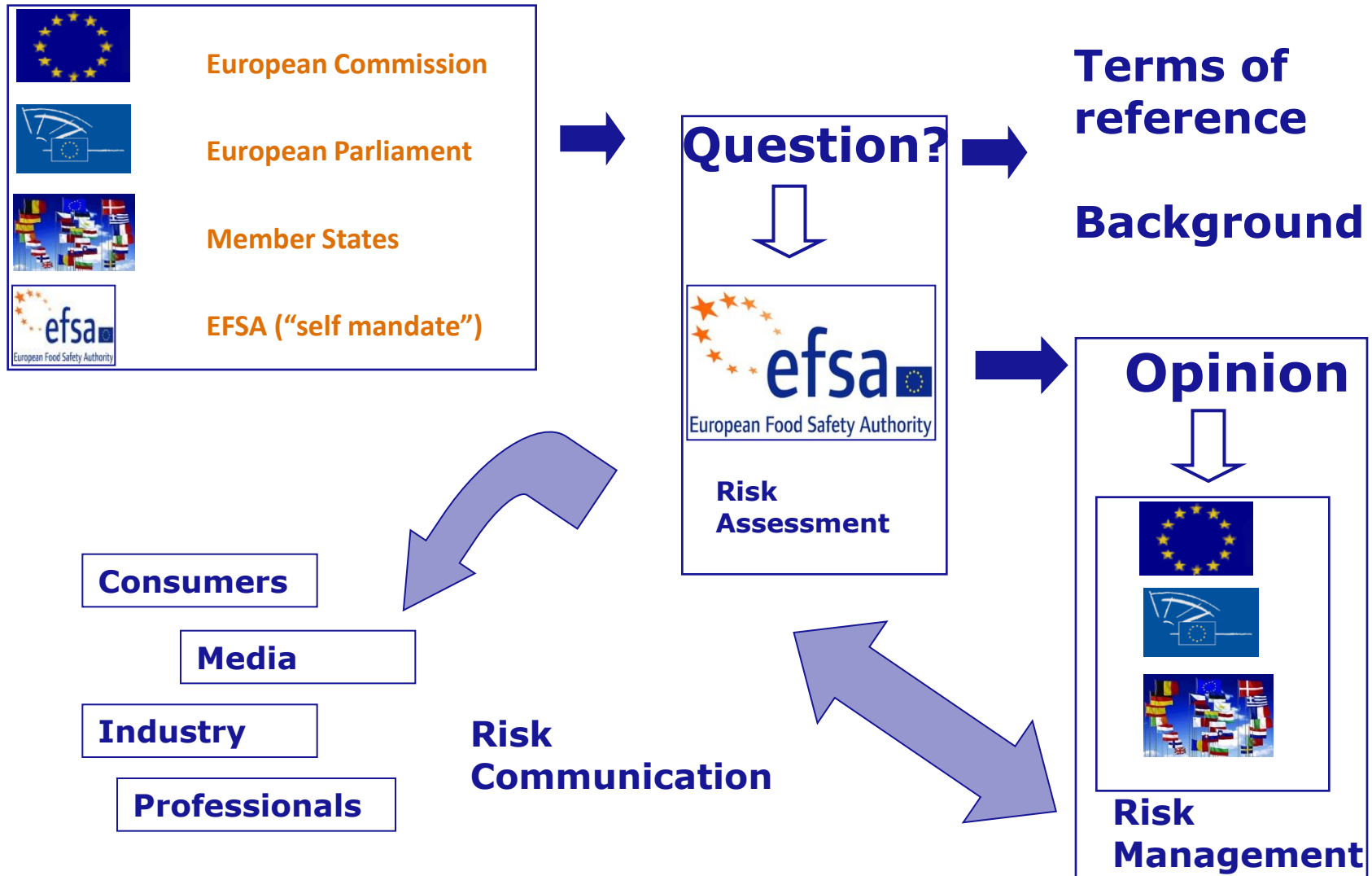
EFSA's Role in Risk Analysis

Methodology Codex Alimentarius:

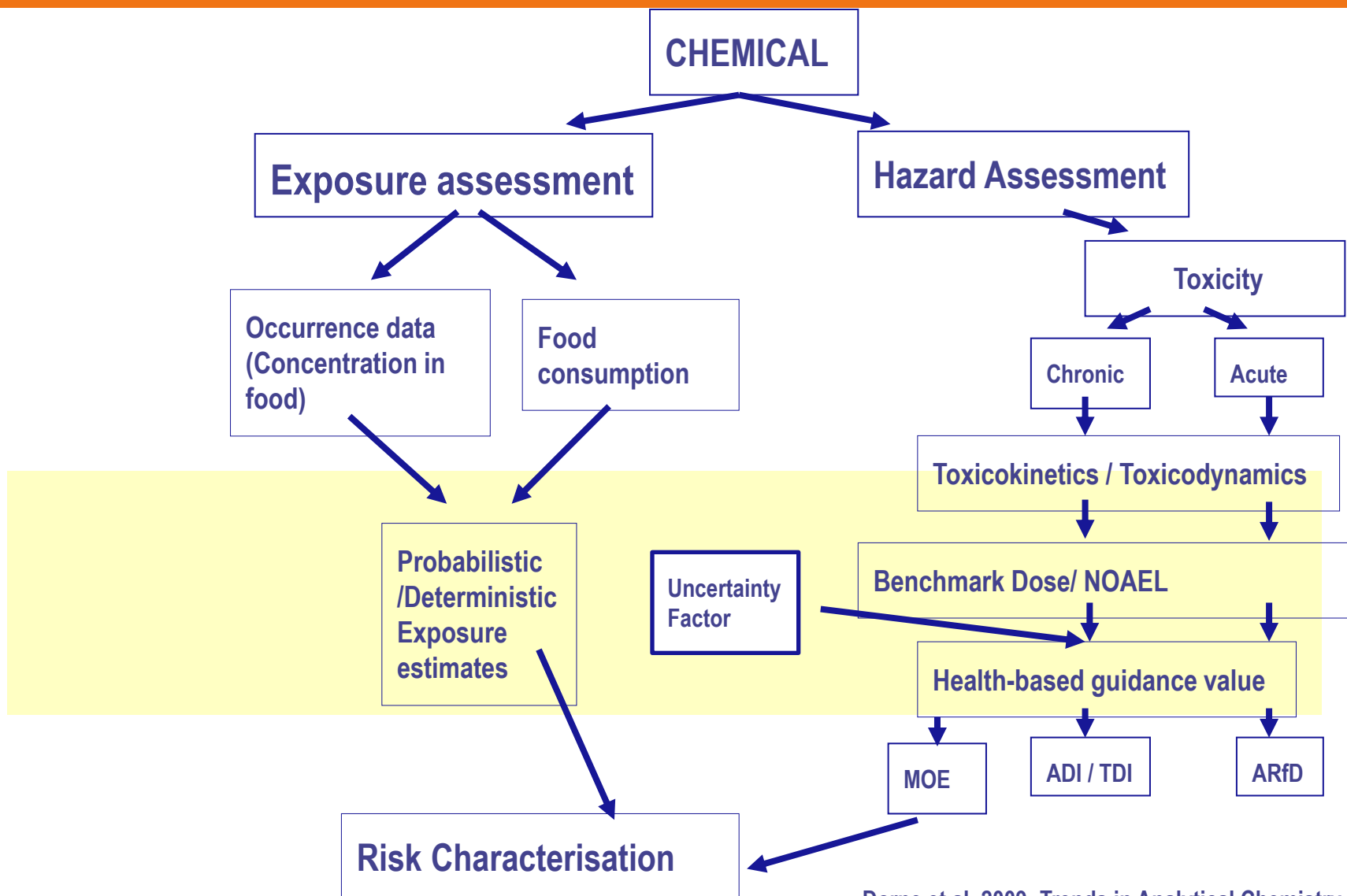
- Hazard identification
- Hazard characterisation
- Exposure assessment
- Risk characterisation



FROM QUESTION TO ANSWER



IN A NUTSHELL...



“All things are toxic and there is nothing without poisonous qualities: it is only the dose which makes something a poison”

PARACELSUS (1493-1541)

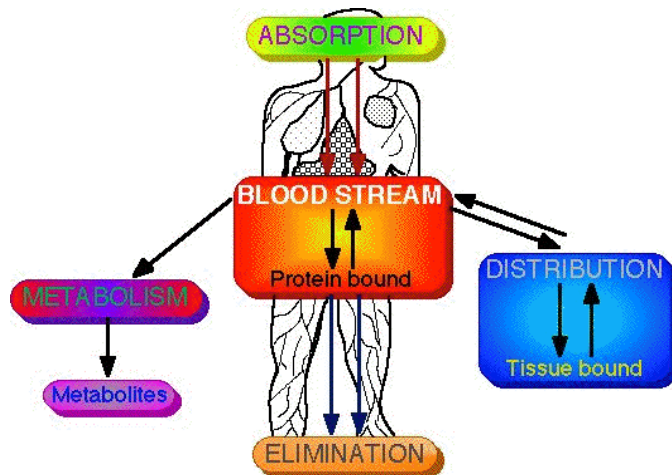


Toxicology

What the body does to a chemical and what a chemical does to the body

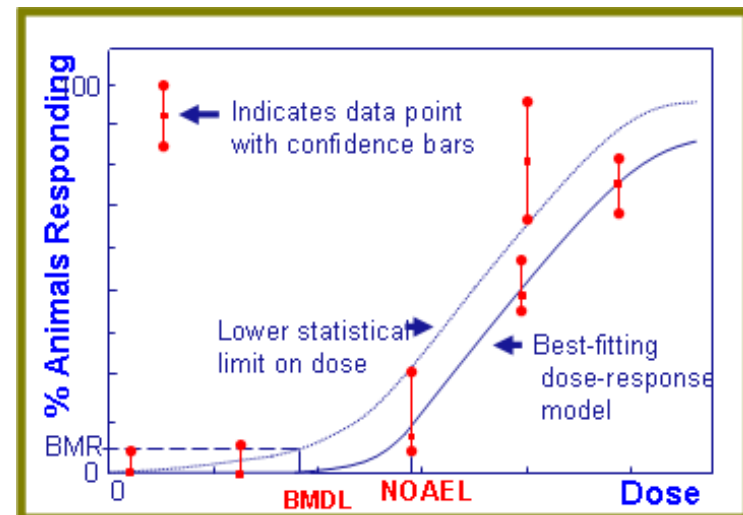
Toxicokinetics

What the body does to a chemical
How the chemical is eliminated from the body or activated into a toxic species (ADME)



Toxicodynamics

What a chemical does to the body
How the chemical exerts its toxicity target receptor/cell/organ



EDITORIAL

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PUBLISHED: 27 March 2015

doi:10.2903/j.efsa.2015.e13031

Increasing robustness, transparency and openness of scientific assessments

Hardy A, Dorne JLCM, Aiassa E, Alexander J, Bottex B, Chaudhry Q, Germini A, Nørrung B, Schlatter J, Verloo D, Robinson T

Scientific assessments are evidenced-based and demand rigorous methodologies to collect, evaluate and integrate scientific evidence, together with transparent and open communication of the processes and results of the assessment. A structured and clearly documented approach is essential if the outcome of the scientific assessment is to be communicated unambiguously to decision makers, the wider scientific community and stakeholders. This will help to clearly focus on key issues and allow reproducibility of the assessments between expert groups and organisations.

Scientific advisory bodies recognise a need to improve the transparency and openness of scientific assessments in line with today's normative and societal expectations. Open scientific assessment can be defined as a decision support process where there is not only full transparency (showing what has

DATA/EVIDENCE AVAILABLE IN CHEMICAL RA

Tier	Exposure Assessment		Hazard identification		Hazard characterisation		Risk Characterisation
	Occurrence	Consumption	TK	TD	TK	TD	
0	Semi-Q	Default values	No data	No data	<i>in silico</i> Read across	Default values TTC Read across <i>In silico</i> Default UF	e.g. Default values Qualitative
1	Point estimates	Point estimates in food categories	<i>In silico</i> Limited data Semi-Q	<i>In silico</i> Limited data Read across	<i>in silico</i> Basic TK Read across	<i>in silico</i> Read across NOAEL Default UF	e.g. Semi-quantitative
2	Measured data	Measured in some food categories	Dossier data Qttve	Dossier Data	<i>in silico</i> ADME data	NOAEL/ BMDL Default <i>in silico</i> UF	e.g. Quantitative Deterministic/ Probabilistic
3	Large measured dataset	Full patterns - food categories	Dossier and/or lit. (<i>in vitro</i> , <i>in vivo</i>)	Data in dossier and/or lit. (<i>in vitro</i> , OMICs, epi)	MoA/AOP, Epi data, PB-PK model, BBDR, BMDL Chemical adjustment (CSAF)	specific factor	e.g. Quantitative Full probabilistic

DATA FROM DOSSIER: REGULATED PRODUCT

Tier	Exposure Assessment		Hazard identification		Hazard characterisation		Risk Characterisation
	Occurrence	Consumption	TK	TD	TK	TD	
2	Measured data	Measured in some food categories	Dossier data Qttve	Dossier Data: genotox, tox	<i>in silico</i> ADME data	NOAEL/ BMDL Default UF	e.g. Quantitative Deterministic/ Probabilistic

DATA-RICH CHEMICAL: CADMIUM

Tier	Exposure Assessment		Hazard identification		Hazard charact		Risk Characterisation
	Occurence	Consumption	TK	TD	TK	TD	
3	Large measured dataset	Full patterns - food categories	Human Data Biomarker excretion/ blood	Human data Biomarker renal toxicity	PB-PK-BMDL (Human data) Chemical specific adjustment factor (CSAF)		e.g. Quantitative Full probabilistic

DATA-POOR: EMERGING MYCOTOXIN

Tier	Exposure Assessment		Hazard identification		Hazard characterisation		Risk Characterisation
	Occurrence	Consumption	TK	TD	TK	TD	
0	Semi-Q	Default values	No data	No data	<i>in silico</i> Read across	Default values TTC Read across <i>In silico</i> Default UF	e.g. Default values Qualitative



OpenFoodTox: EFSA's Open Source Hazards Database

OPENFOODTOX: EFSA' S CHEMICAL HAZARDS DATABASE

- **Catalogue of EFSA's chemical toxicity data since creation**
 - Contaminants (Human and Animal health)
 - Vitamins and minerals (Human health) (NDA),
 - Food additives and Nutrient Sources, Food contact materials, Flavourings and processing aids (Human Health)
 - Feed Additives (Human and Animal Health, Ecotoxicology)
 - Pesticides (Human and Animal health, Ecotoxicology)

- **Easy Reference and Crisis**
 - One reference DB Chemical Hazards: Search easily and efficiently
 - Crisis: Quick and Easy access to all EFSA's Hazard Data

- **International Harmonisation**
 - Use OECD Harmonised Templates (OHT) for data model (ECHA/OECD) compatible with IUCLID/ ECHA-OECD QSAR toolbox
 - Search compounds by name, CAS number on e-chem portal
 - Generate data sheet as summary of hazard id and charact (June 2016)

WHAT DOES OPENFOODTOX CONTAIN ?

○ Chemical Information

Information on chemical nomenclature (EU nomenclature, IUPAC, CAS...), trade name, chemical group/panel (i.e. pesticide), chemical use (i.e. fungicide), chemical structure (i.e. triazoles, organophosphates...).

○ Document descriptors

Information on EFSA's opinion for the specific chemical or group of chemicals. Info from EFSA 's RAW system (question number, mandate, number), link to the document

○ Toxicity Endpoint/ Hazard identification

Information on critical toxicity study using OECD picklists when possible (species, dose, target organ...)

○ Critical study to demonstrate genotoxicity status

Providing essential information of critical genotoxicity study when assessed

○ Hazard /Risk characterisation

Information for health based guidance values (ADI/TDI) uncertainty factors...

TOWARDS AN OPENSOURCE HAZARD DATABASE

Tier	Exposure Assessment		Hazard identification		Hazard characterisation		Risk Characterisation
	Occurrence	Consumption	TK	TD	TK	TD	
0	Semi-Q	Default values	No data	No data	<i>in silico</i> Read across	Default values TTC In silico	e.g. Default values Qualitative
1	Point estimates	Point estimates in food categories	<i>In silico</i> Limited data Semi-Q	In silico data Read across	<i>in silico</i> Basic TK Read across	in silico NOAEL Default UF	e.g. Semi-quantitative
2	Measured data	Measured in some food categories	Dossier data Qttve	Dossier Data	<i>in silico</i> ADME data	NOAEL/ BMDL Default in silico UF	e.g. Quantitative Deterministic/ Probabilistic
3	Large measured dataset	Full patterns - food categories	Dossier and/or lit. (<i>in vitro</i> , <i>in vivo</i>)	Dossier and/or lit. (<i>in vitro</i>, OMICs, Epi)	MoA/AOP, Epi data, PB-PK model, BBDR, BMDL, CSAF		e.g. Quantitative Full probabilistic

CONTENT

Microstrategy Tool: <https://dwh.efsa.europa.eu/bi/asp/Main.aspx?rwtrep=400>

Full Download Knowledge junction:

https://zenodo.org/record/344883#.WUDqK_mGPIU

1,650
Scientific
outputs
(metadata +
DOI)

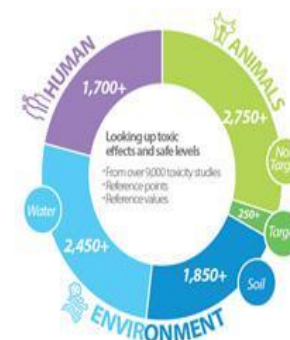
4,400
Substances
(chemical
identifiers
including
SMILES)

10,000
Toxicologica
I endpoint
studies

140 Positive
genotoxicity
studies

12,000 risk
assessment
summaries

Use
OpenFoodTox
for



FILTER

Substance Browser

Sub Name (4352)

(All)

- (-)-3,7-Dimethyl-6-octen-1-ol
- (-)-Alpha-cedrene
- (-)-Alpha-elemol
- (-)-alpha-Santalene
- (-)-Bornyl acetate
- (-)-Hyoscyamine and (-)-Scopolamine group
- (-)-Rhodinol
- (-)-Scopolamine
- ((E)-2-(2-hydroxymethylphenyl)-2-methylpropane
- ((2-((2-Ethyl-6-methylphenyl)oxy)ethyl)oxy)ethane
- ((2-((Ethoxymethyl)(2-ethyl-1-hydroxyethyl)oxy)ethyl)oxy)ethane
- (+)-Alpha-cedrene
- (1R,2R,5S)-5-Isopropenyl-2-methyl-6-octen-1-ol
- (1R,2R,5S)-Isodihydrocarveol
- (1R,2S,5R)-5-methyl-2-(1-methylethoxy)ethane
- (1R,2S,5R)-Menthyl hexanoate
- (1R,2S,5R)-Menthyl salicylate
- (1R,2S,5R)-N-(2-(Pyridine-2-yl)ethyl)-3-piperidinecarboxamide
- (1R,2S,5R)-N-(4-Methoxyphenyl)-2-methyl-6-octen-1-ol
- (1R,2S,5R)-N-(Ethoxycarbonyl)-2-methyl-6-octen-1-ol
- (1R,2S,5R)-N-Cyclopropyl-2-methyl-6-octen-1-ol
- (1R,2S,5R)-N,N-Dimethyl-2-methyl-6-octen-1-ol
- (1R,2S,5S)-3-Methoxy-2-methyl-6-octen-1-ol
- (1R,2S,5S)-neo-Dihydrocarveol

Substance Characterisation

Substance	has	Component	CAS Number	EC Ref No	Molec formula
(-)-3,7-Dimethyl-6-octen-1-ol	as such	(-)-3,7-Dimethyl-6-octen-1-ol	7540-51-4	231-415-7	C10H20
(-)-Alpha-cedrene	as such	(-)-Alpha-cedrene	469-61-4	207-418-4	C15H24

EFSA outputs

Substance	Output Id	Legal Basis	Panel	Published	Title
(-)-3,7-Dimethyl-6-octen-1-ol	2180	Commission Regulation (EC)	EFSA CEF	02/20/2013	Scientific Opinion on Flavouring Group Revision 4 (FGE.06Rev4): Straight- and branched-chain aliphatic aldehydes

Hazard Characterisation: Reference points

Substance	Author	Year	Output ID	Study	Test Type	Species	Route	Duration (days)
(-)-Hyoscyamine and (-)-Scopolamine group	EFSA CONTAM	2013	2396	Human health	study with volunteers	Human	Not reported	
(1R,2S,5R)-N-(2-(Pyridine-2-yl)ethyl)-3-piperidinecarboxamide	EFSA CEF	2014	2524	Human health	subchronic	Rat	oral: feed	90

Hazard Characterisation: Reference values

Substance	Author	Year	Output Id	Assessment	qualifier	value	unit
(-)-3,7-Dimethyl-6-octen-1-ol	EFSA CEF	2013	2180	TTC Cramer Class I	=	30	µg/kg bw/day
(-)-Alpha-elemol	EFSA AFC	2006	2232	TTC Cramer Class I	=	30	µg/kg bw/day

Genotoxicity

Substance	Author	Year	Output Id	Genotoxicity
9,10-Dihydroxy stearic acid oligomers	EFSA AFC	2003	344	Not detected
Acrylic acid, methyl ester, telomer with 1-dodecanethiol, C16-C18 alkyl esters	EFSA AFC	2003	344	Negative

Open Source in silico Tools to Quantify Toxicological Processes



OPENFOODTOX AND *IN SILICO* TOOLS

- Case studies to develop *in silico* tools
 - QSAR model on pesticide Toxicity in bees (OpenFoodTox, US-EPA, DEMETRA DB) : Classifier
 - QSAR model to predict LC₅₀ in rainbow trout (OpenFoodTox) : Continuous model
 - Physico-chem properties, structure, toxicity : $R^2 > 0.75$
 - QSAR model to predict NOAEL in rats (OpenFoodTox, Fraunhofer) : Continuous model
 - Physico-chem properties, structure, toxicity : $R^2 > 0.75$
- Scientific report Summer 2017
 - QSAR model to predict NOAEL for liver toxicity in rats (OpenFoodTox, Fraunhofer) : Continuous model



ELSEVIER

Contents lists available at ScienceDirect

Environmental Toxicology and Pharmacology

journal homepage: www.elsevier.com/locate/etap



QSAR models for predicting acute toxicity of pesticides in rainbow trout using the CORAL software and EFSA's OpenFoodTox database



Andrey A. Toropov^a, Alla P. Toropova^{a,*}, Marco Marzo^a, Jean Lou Dorne^b, Nikolaos Georgiadis^b, Emilio Benfenati^a

^a Department of Environmental Health Science, Laboratory of Environmental Chemistry and Toxicology, IRCCS-Istituto di Ricerche Farmacologiche Mario Negri, Via La Masa 19, 20156 Milano, Italy

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ARTICLE INFO

Keywords:

QSAR
OECD
Monte Carlo method
Rainbow trout
Toxicity
CORAL software

ABSTRACT

Optimal (flexible) descriptors were used to establish quantitative structure – activity relationships (QSAR) for toxicity of pesticides (n = 116) towards rainbow trout. A heterogeneous set of hundreds of pesticides has been used, taken from the EFSA's chemical Hazards Database: OpenFoodTox. Optimal descriptors are preparing from simplified molecular input-line entry system (SMILES). So-called, correlation weights of different fragments of SMILES are calculating by the Monte Carlo optimization procedure where correlation coefficient between endpoint and optimal descriptor plays role of the target function. Having maximum of the correlation coefficient for the training set, one can suggest that the optimal descriptor calculated with these correlation weights can correlate with endpoint for external validation set. This approach was checked up with three different distributions into the training ($\approx 85\%$) set and external validation ($\approx 15\%$) set. The statistical characteristics of these models are (i) for training set correlation coefficient (r^2) ranges 0.72–0.81, and root mean squared error (RMSE) ranges 0.54–1.25; (ii) for external (validation) set r^2 ranges 0.74–0.84; and RMSE ranges 0.64–0.75. Computational experiments have shown that presence of chlorine, fluorine, sulfur, and aromatic fragments is promoter of increase for the toxicity.

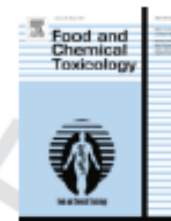


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Contents lists available at ScienceDirect

Food and Chemical Toxicology

journal homepage: www.elsevier.com



The application of new HARD-descriptor available from the CORAL software to building up NOAEL models

Alla P. Toropova^{a,*}, Andrey A. Toropov^a, Marco Marzo^a, Sylvia E. Escher^b, Jean Lou Dorne^c, Nikolaos Georgiadis^c, Emilio Benfenati^a

^a Department of Environmental Health Science, Laboratory of Environmental Chemistry and Toxicology, IRCCS-Istituto di Ricerche Farmacologiche Mario Negri, Via La Masa 19, 20156 Milano, Italy

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NOAEL

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Monte Carlo method

CORAL software

ABSTRACT

Continuous QSAR models have been developed and validated for the prediction of no-observed-adverse-effect (NOAEL) in rats, using training and test sets from the Fraunhofer RepDose® database and EFSA's Chemical Hazards Database: OpenFoodTox. This paper demonstrates that the HARD index, as an integrated attribute of SMILES, improves the prediction power of NOAEL values using the continuous QSAR models and Monte Carlo simulations. The HARD-index is a line of eleven symbols, which represents the presence, or absence of eight chemical elements (nitrogen, oxygen, sulfur, phosphorus, fluorine, chlorine, bromine, and iodine) and different kinds of chemical bonds (double bond, triple bond, and stereo chemical bond). Optimal molecular descriptors calculated with the Monte Carlo technique (maximization of correlation coefficient between the descriptor and endpoint) give satisfactory predictive models for NOAEL. Optimal molecular descriptors calculated in this way with the Monte Carlo technique (maximization of correlation coefficient between the descriptor and endpoint) give amongst the best results available in the literature. The models are built up in accordance with OECD principles.



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QSAR models for predicting acute toxicity of pesticides in rainbow trout using the CORAL software and EFSA's OpenFoodTox database

Andrey A. Toropov^a, Alla P. Toropova^{a,*}, Marco Marzo^a, Jean Lou Dorne^b, Nikolettina Benfenati^a,
Emilio Benfenati^a^a Department of Environmental Health Science, Laboratory of Environmental Chemistry and Toxicology, IRCCS-Istituto di Ricerche Farmacologiche "Mario Negri", Via Sallustiana 155, 00100 Roma, Italy^b Scientific Committee and Emerging Risks Unit, European Food Safety Authority, Via Carlo Magno 1A, 43126 Parma, Italy

ARTICLE INFO

Keywords:
QSAR

ABSTRACT

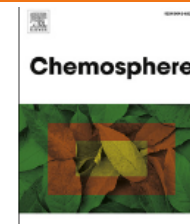
Optimal (flexible) descriptors were used to establish quantitative structure–activity relationships (QSAR) for the prediction of the acute toxicity of pesticides (n = 116) towards rainbow trout. A hetero-



Contents lists available at [ScienceDirect](#)

Chemosphere

journal homepage: www.elsevier.com/locate/chemosphere



Predicting acute contact toxicity of pesticides in honeybees (*Apis mellifera*) through a k-nearest neighbor model



F. Como ^{a,*}, E. Carnesecchi ^b, S. Volani ^b, J.L. Dorne ^b, J. Richardson ^b, A. Bassan ^c,
M. Pavan ^c, E. Benfenati ^a

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^c S-IN Soluzioni Informatiche S.r.l., via G. Ferrari 14, 36100 Vicenza, Italy

H I G H L I G H T S

- A model to predict acute contact toxicity for bees was built for screening pesticides.
 - The model developed will address future risk assessments of pesticides of concern.
 - The accuracy of k-NN model is good and equal to 65% for the highly toxic compounds.
-

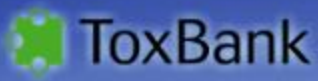
VEGA

TRY THE NEW TOXREAD

DOWNLOAD VEGA

ABOUT QSAR

NEWS



Our Vision

Our Mission

News & Updates

June, 2016 Working on behalf of UBA The Platform for prioritising possible persistent, bioaccumulative and toxic (PBT) chemicals relies on VEGA
June, 2016

On site

June, 2016 How to interpret VEGA n explanation on how to has been added
June, 2016 An example on how to us

Our Community

SCIENTIFIC REPORT OF EFSA

Modern methodologies and tools for human hazard assessment of chemicals¹

European Food Safety Authority^{2,3}

European Food Safety Authority (EFSA), Parma, Italy

*This scientific output, published on 11 July 2014, replaces the earlier version published on 24 April 2014**

ABSTRACT

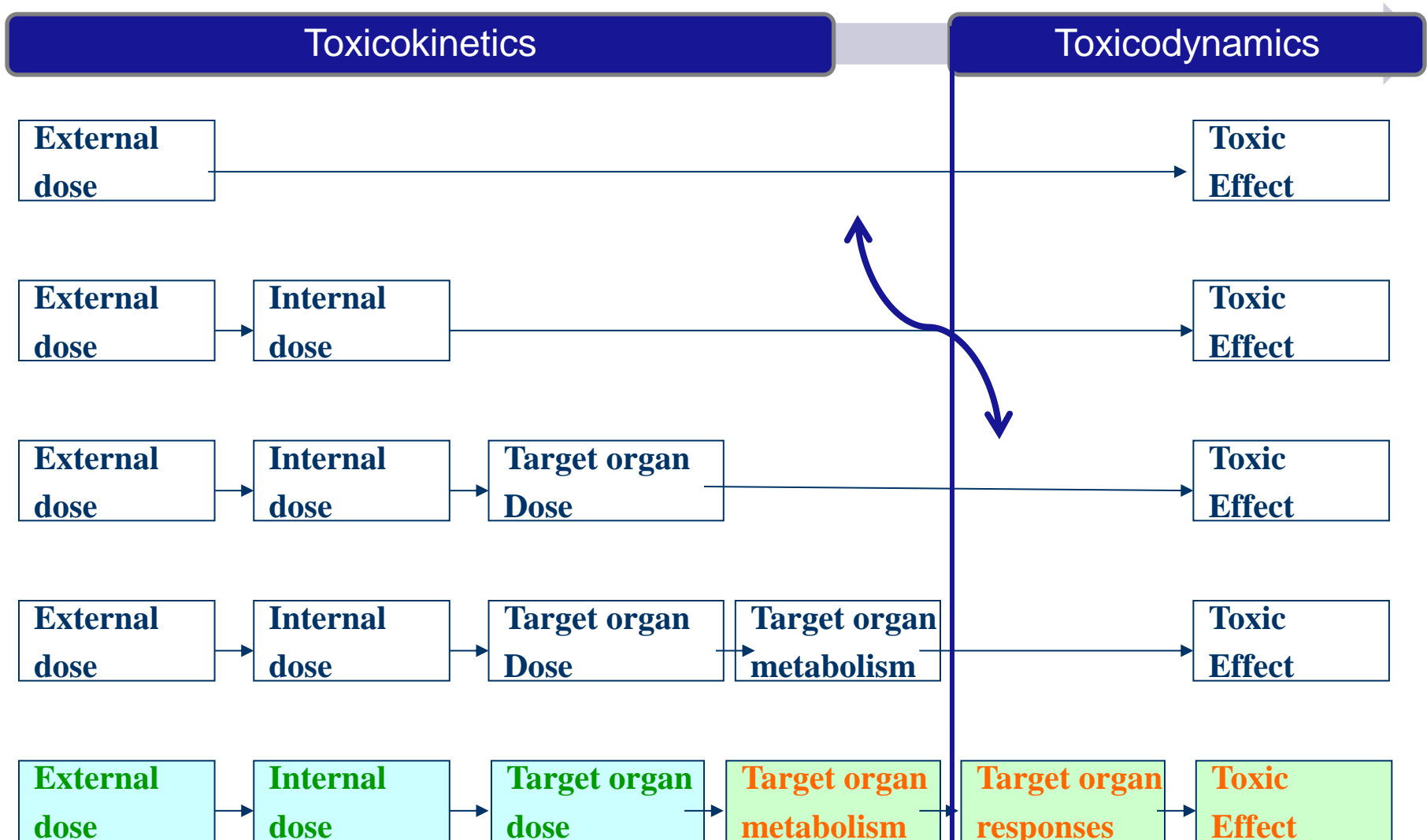
This scientific report provides a review of modern methodologies and tools to depict toxicokinetic and toxicodynamic processes and their application for the human hazard assessment of chemicals. The application of these methods is illustrated with examples drawn from the literature and international efforts in the field. First, the concepts of mode of action/adverse outcome pathway are discussed together with their associated terminology and recent international developments dealing with human hazard assessment of chemicals. Then modern methodologies and tools are presented including *in vitro* systems, physiologically-based models, *in silico* tools and OMICs technologies at the level of DNA/RNA (transcriptomics), proteins (proteomics) and the whole metabolome (metabolomics). Future perspectives for the potential applications of these modern methodologies and tools in the context of prioritisation of chemicals, integrated test strategies and the future of risk assessment are discussed. The report concludes with recommendations for future work and research formulated from consultations of EFSA staff, expert Panels and other international organisations.

© European Food Safety Authority, 2014

KEY WORDS

mode of action, adverse outcome pathway, integrated testing strategy, physiologically-based models, *in silico*, OMICs

-Levels of Knowledge, Toxicokinetic and Toxicodynamic processes



New Data requirements for pesticides Regulation 283- 284/2013 : TK Data

In vivo TK studies in animals

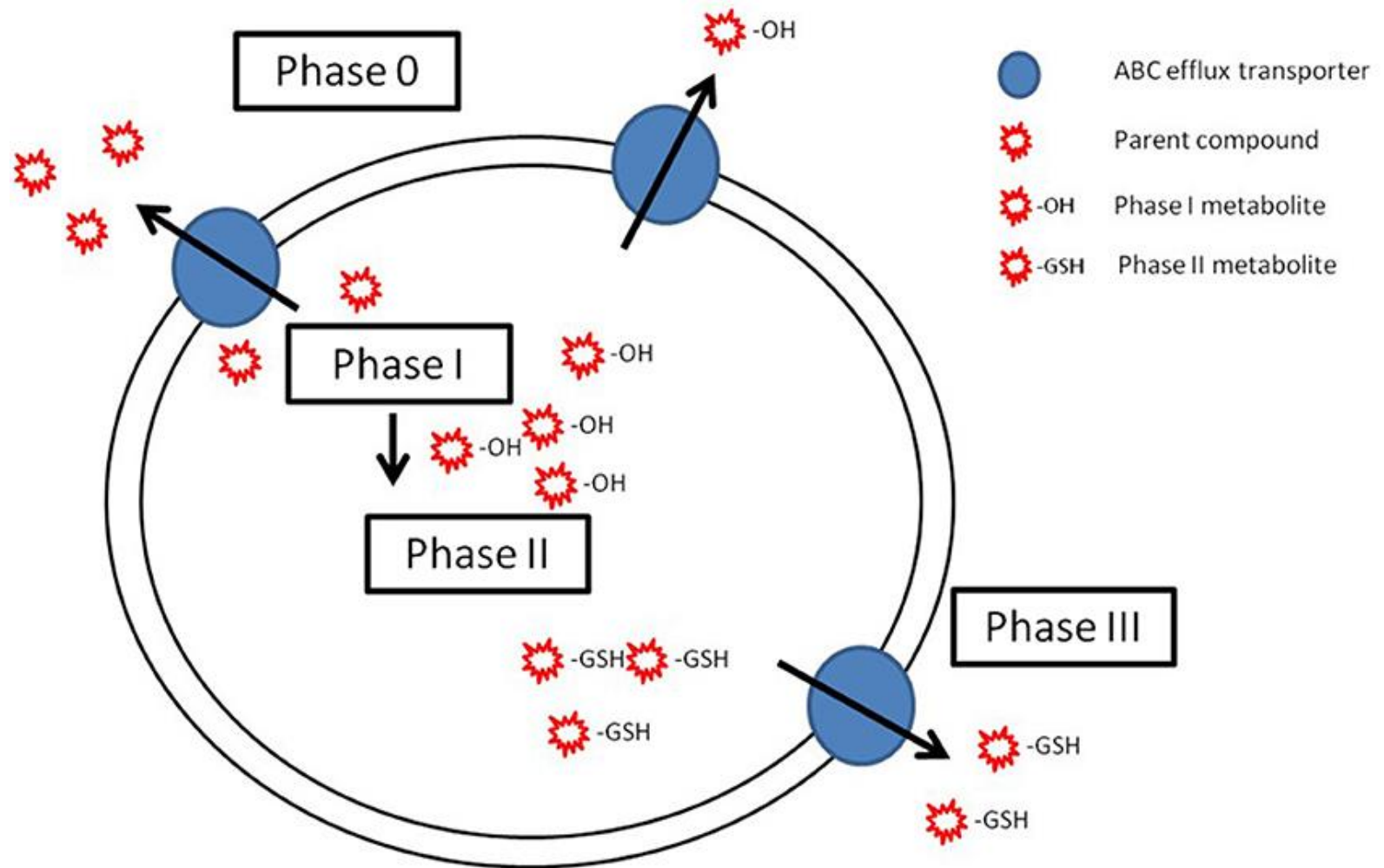
- ✓ Blood/ tissues [C] for active substance/relevant metabolites (C_{\max} ; AUC)
in relevant species understanding toxicity studies
- ✓ Investigating entero-hepatic circulation

Comparative Animal versus human microsomes or intact cell systems

Relevance animal tox -guide interpretation, further define testing strategy
e.g. human *in vitro* metabolite not in test species

- ✓ Protocols available publicly incl. ECVAM work on developing TK standards
- ✓ *In vitro* models hepatic/ non-hepatic microsomes (e.g. intestinal)
- ✓ Major human metabolites (>10% of AD) not at sufficient levels in animal studies further investigated for their toxicity profile.

MAJOR METABOLIC/EXCRETION ROUTES IN HUMANS



MAJOR METABOLIC/EXCRETION ROUTES IN HUMANS

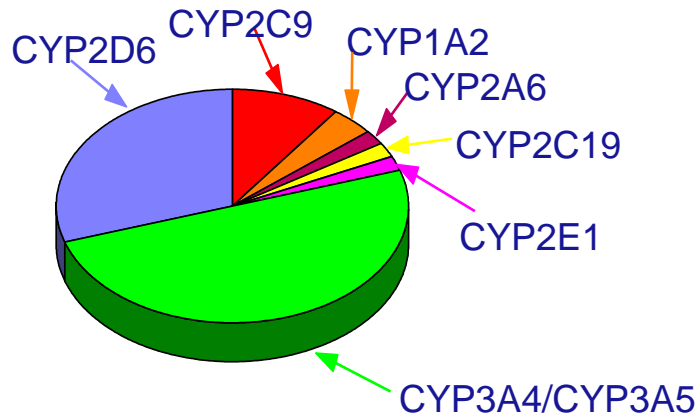
Phase I enzymes
 Cytochrome P-450,
 ADH, Esterases...

Phase II enzymes
 Conjugation reactions

UDP-Glucuronyltransferases,
 Sulphotransferases
 Glutathione-s-transferases
 Methyl-transferases
 N-acetyltransferases
 Amino acid conjugation

Transporters
 Phase 0- Uptake transporters:
 e.g OATPs, OCTs.

Phase III-Efflux pumps:
 e.g ABCs (P-glycoproteins and
 MRPs)



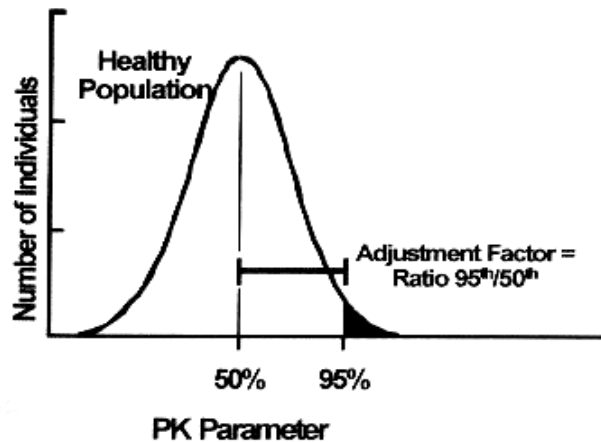
Renal excretion

-HUMAN VARIABILITY IN TOXICOKINETICS-

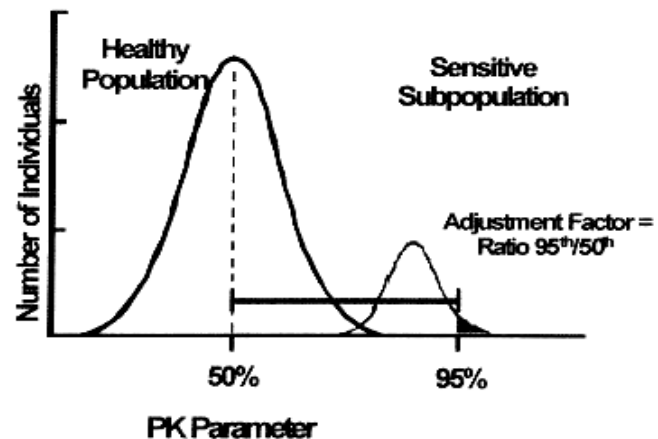
From pharmaceutical database and compounds relevant to food safety,

- ✓ *Identify Phase 0, 1, 2, 3 isoforms in vitro , excretion data etc.*
- ✓ *PK parameters of acute and chronic exposure: Meta-analysis*
- ✓ *Human variability distributions -isoform specific for different subgroups of the population.*

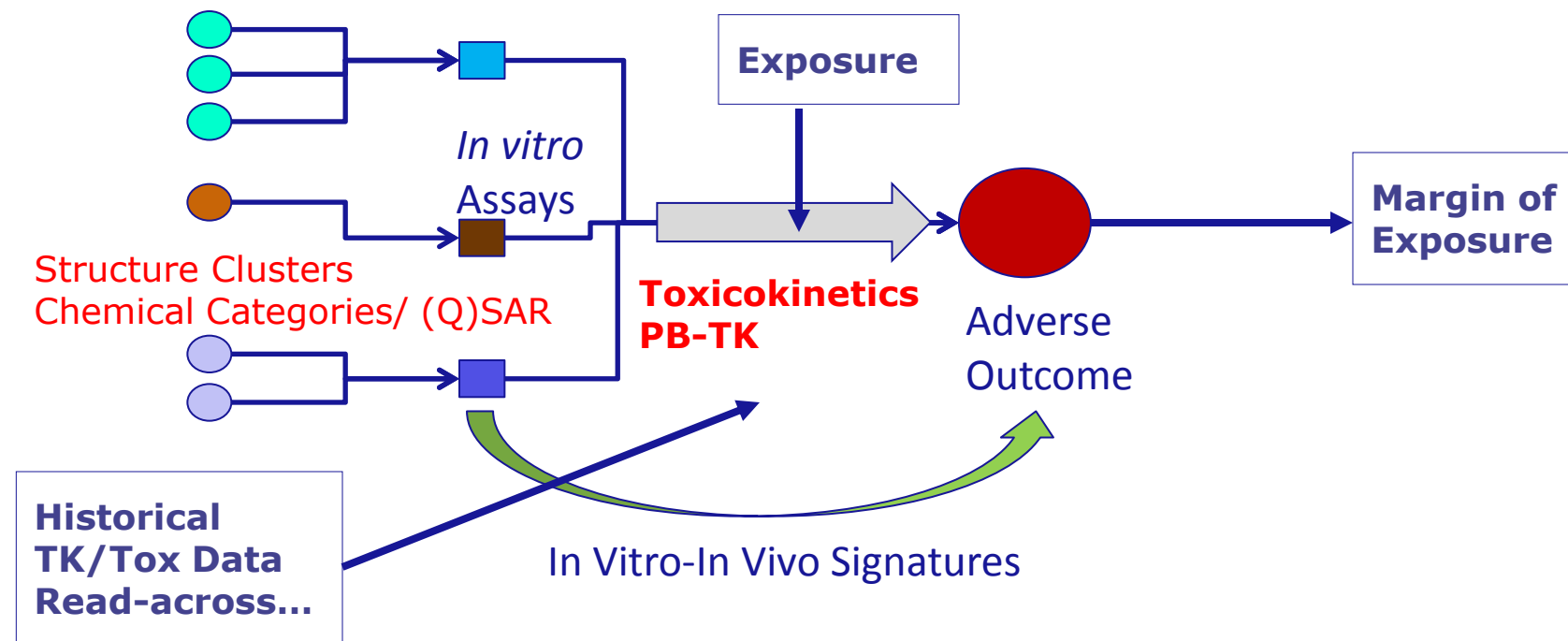
✓ **Unimodal Population**



Bimodal Population



TK AND INTEGRATED TESTING STRATEGIES



Toxicokinetics

In vitro id isoforms phase I, II, transporters.

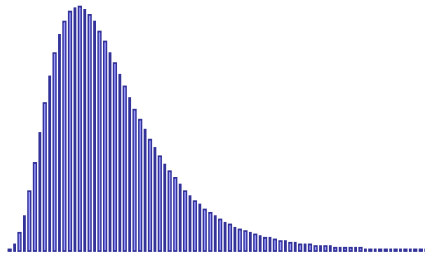
Consequences of metabolism id of toxic moiety(ies)

TK parameters (V_m , K_m , Cl_{int} , F_u).

Use human Variability in TK from historical databases and software

IVIVE

-HUMAN VARIABILITY IN TOXICOKINETICS AND UF-

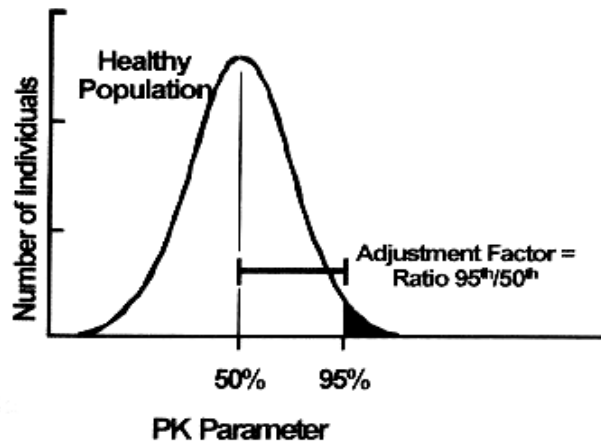


From pharmaceutical database human variability in TK available for many drugs /enzyme isoforms in different subgroups of the population.

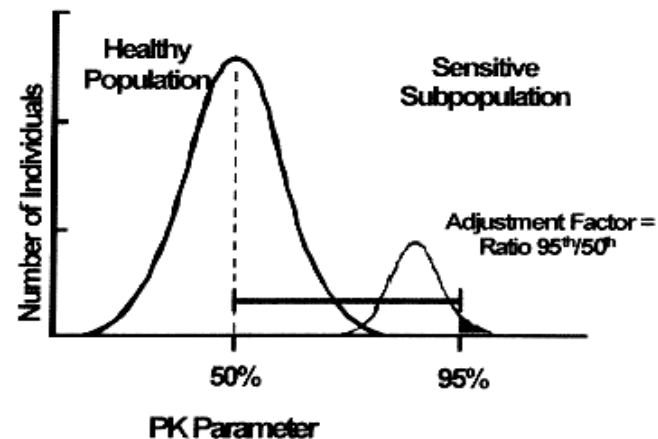
Rationale for meta-analysis of TK data to derive pathway-related uncertainty factors- default distributions.

Can be refined in the future

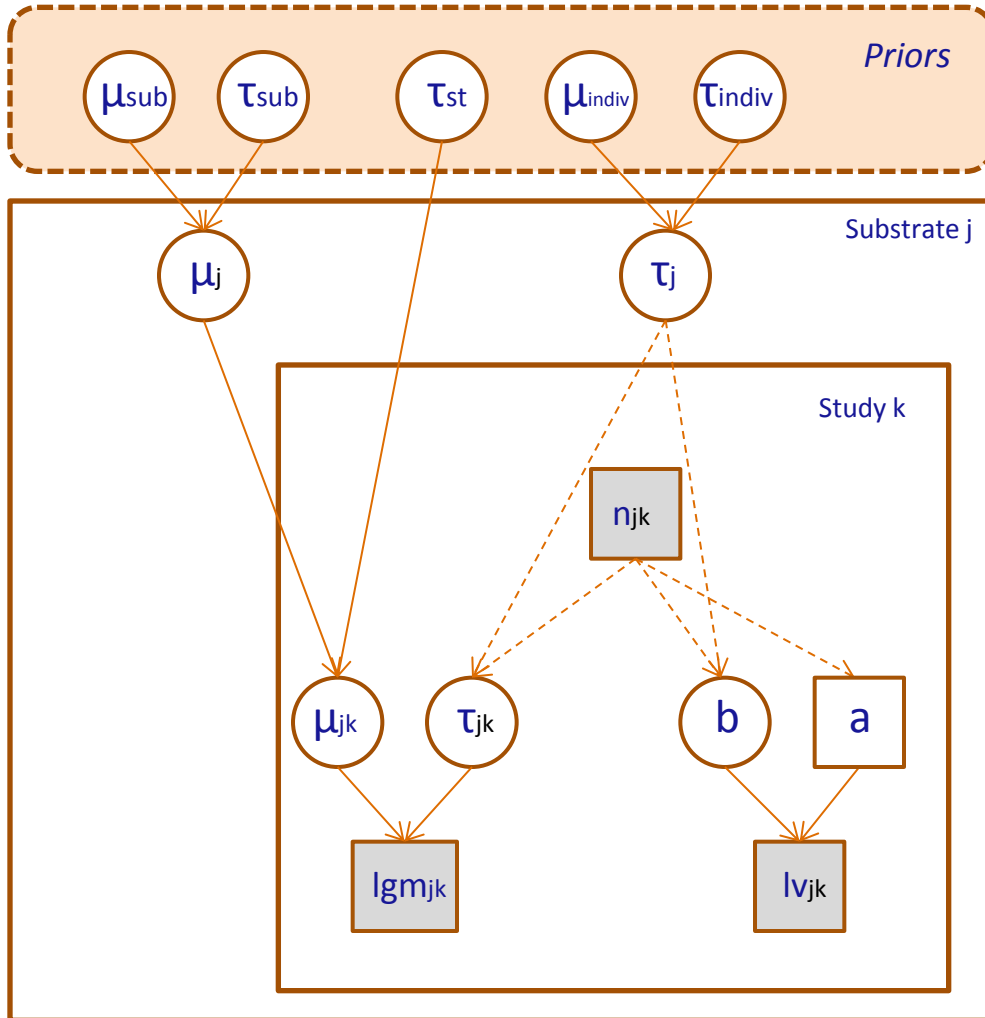
Unimodal Population



Bimodal Population



Bayesian Meta-analysis of TK data



3 levels :inter-study, inter-substrate et inter-individual Variability

$$\tau_j \sim Normal(\mu_{ind}, \frac{1}{\tau_{ind}})$$

$$\mu_j \sim Normal(\mu_{sub}, \frac{1}{\tau_{sub}})$$

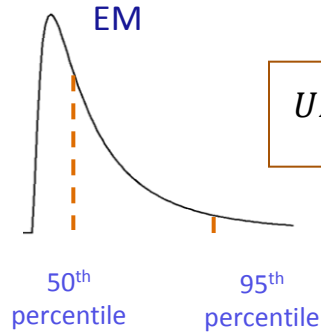


$$\mu_{jk} \sim Normal(\mu_j, \frac{1}{\tau_{st}})$$



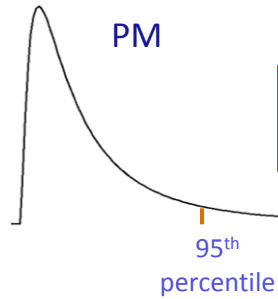
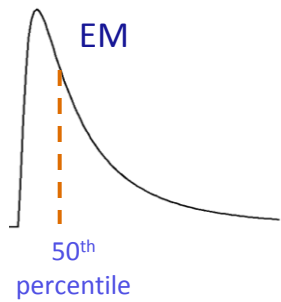
$$lgm_{jk} \sim Normal(\mu_{jk}, \frac{1}{n_{jk}\tau_j})$$

$$lv_{jk} \sim \frac{1}{n_{jk}\tau_j} Chi^2(n_{jk} - 1)$$

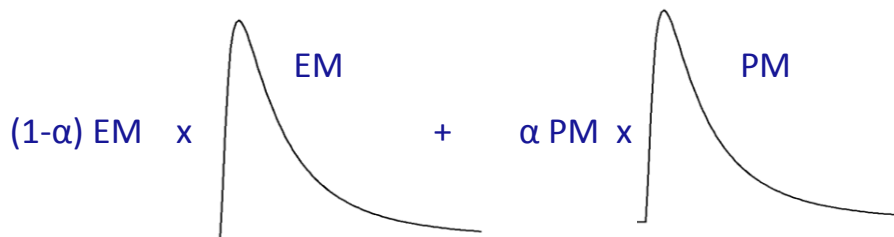


$$UF_{EM} = \frac{p_{95_{EM}}}{p_{50_{EM}}}$$

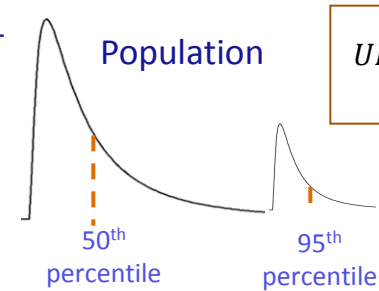
Polymorphic CYP2D6 Example



$$UF_{PM} = \frac{p_{95_{PM}}}{p_{50_{EM}}}$$

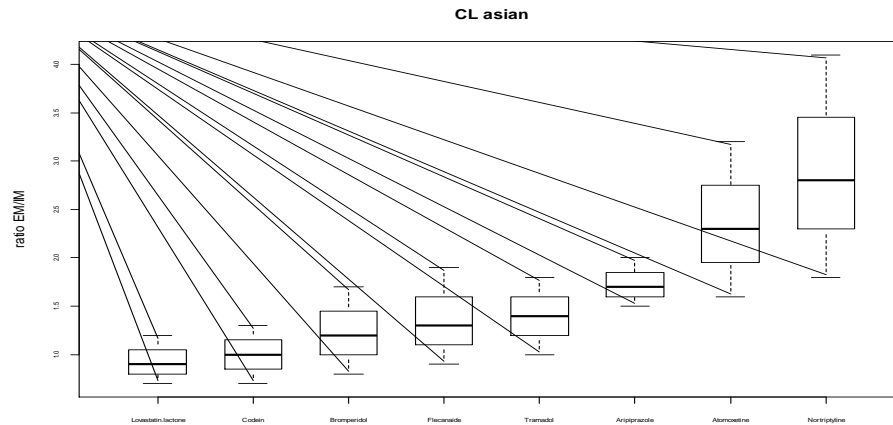
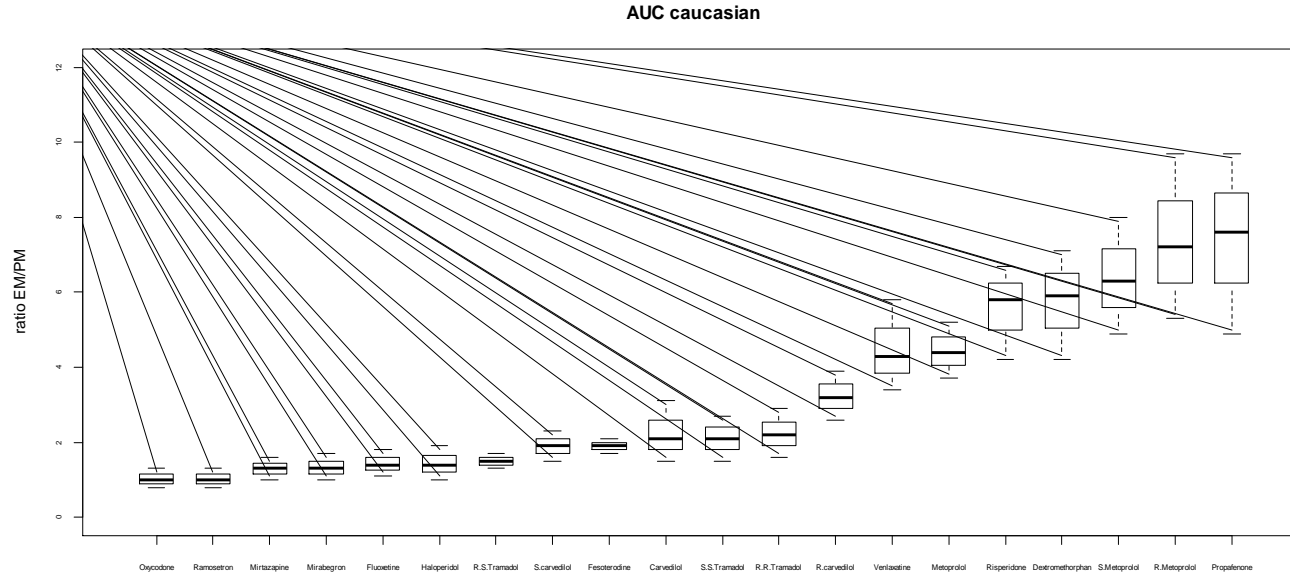


Monte-Carlo
→



$$UF_{pop} = \frac{p_{95_{pop}}}{p_{50_{pop}}}$$

Polymorphic CYP2D6 Example

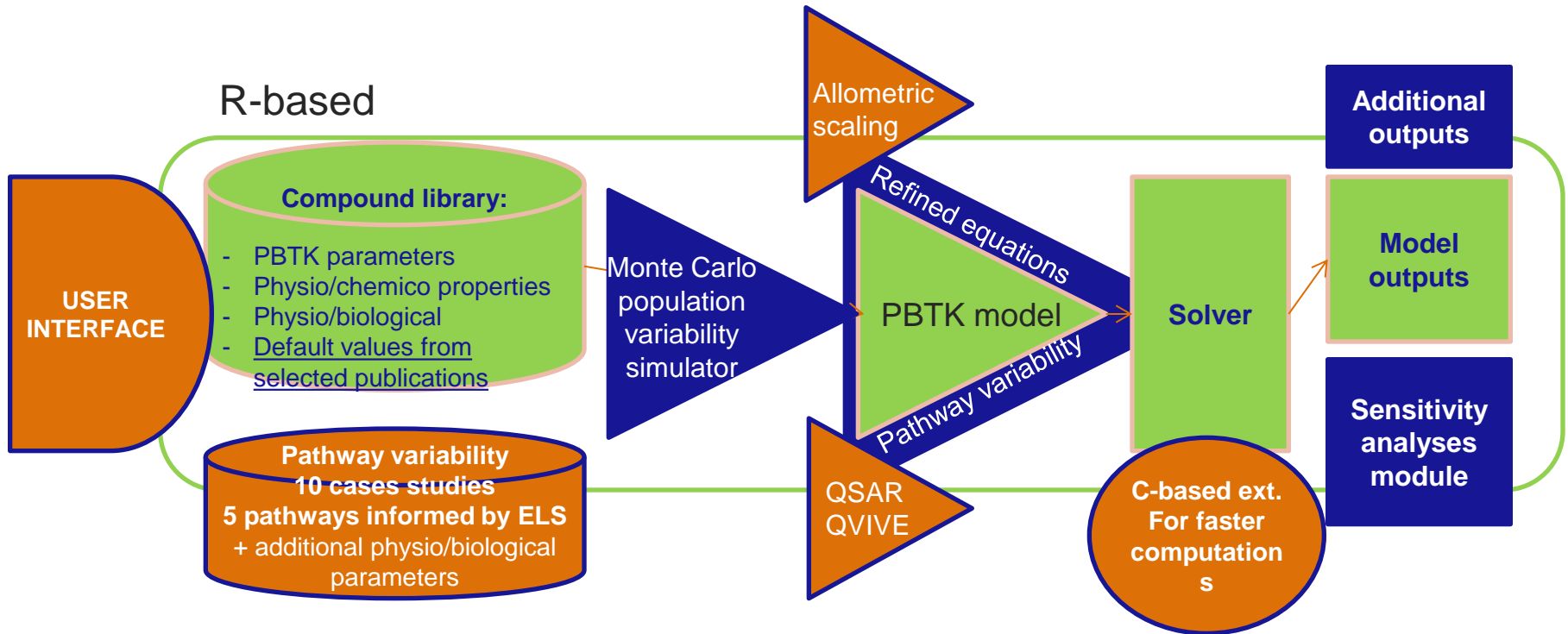


% Dose metabolised by CYP2D6 and differences between Extensive and poor metabolisers in Caucasian and Asian populations

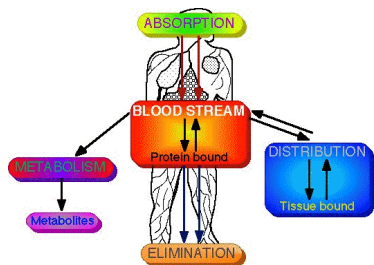
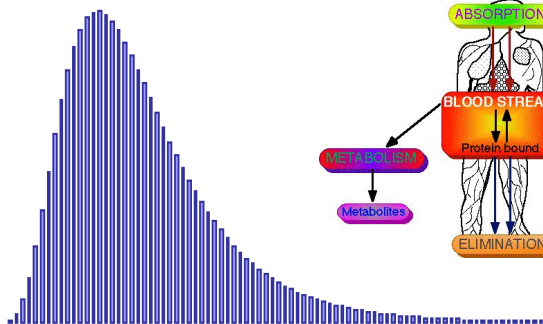
OPEN SOURCE TK MODELS: DATA AND MODELS

- ✓ Collection data physio/ biological param- calibrate TK tools
 - Body weight, variability enzymes expression Gut/liver etc...
 - Human Variability metabolism (CYP isoforms) and excretion using Pharmaceutical DB
 - TK tools from one compartment to multi compartment/PB-PK e.g. blood/liver/gut/kidney
 - Case studies 10 compounds relevant to food and feed safety combining TK and TD: regulated, contaminants
- ✓ In Future Open TK tools in R (spring 2018)
- ✓ In Parallel, TK tools for 5 veterinary species (cow, pig, cat, chicken etc..) and ERA (zebra, trout, earth worm)

Prototype TK Modelling Platform

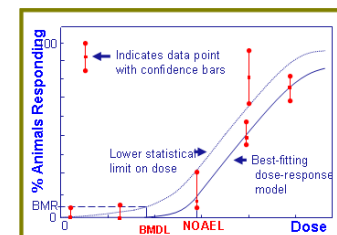


COMBINING VARIABILITY IN TK AND IN VITRO DATA : OPENSOURCE PLATFORM



TK

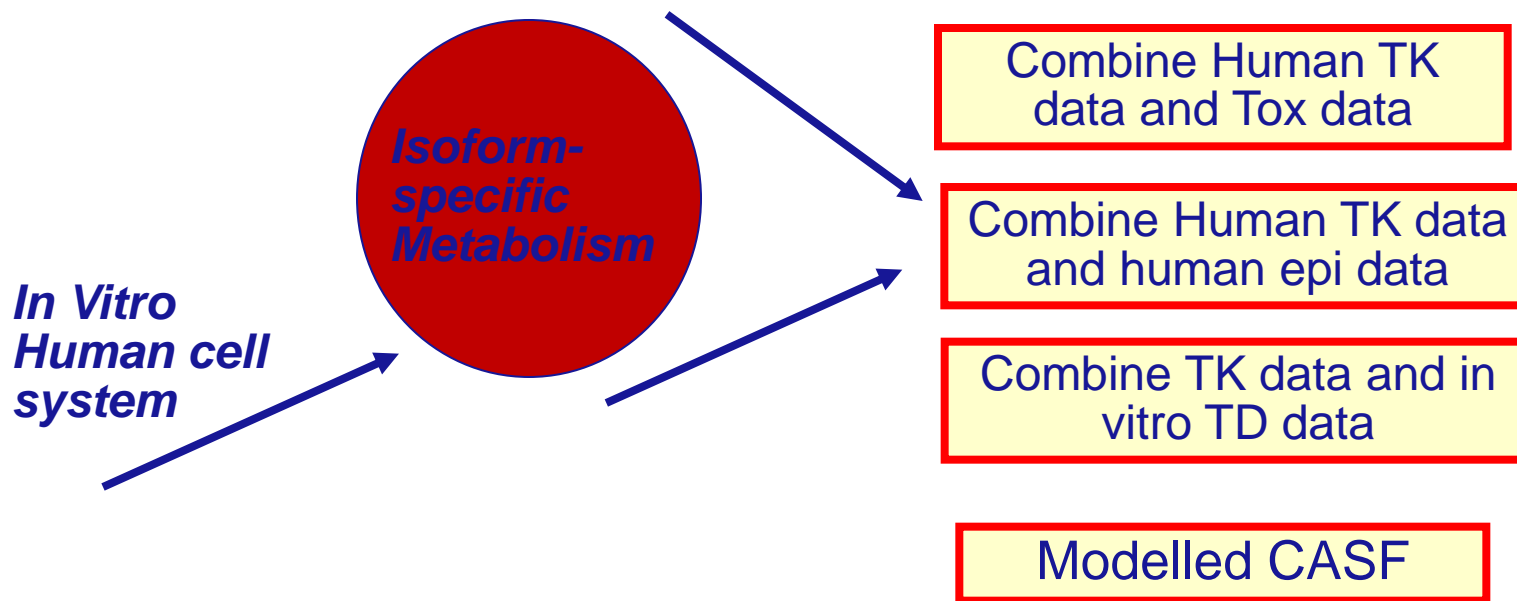
TD



Isoform-specific Variability Distribution : Open Source Tool

Meta-analysis TK studies (acute, chronic) and TD studies (vivo, vitro, epidemio)

Phase I and Phase II enzymes and Transporters



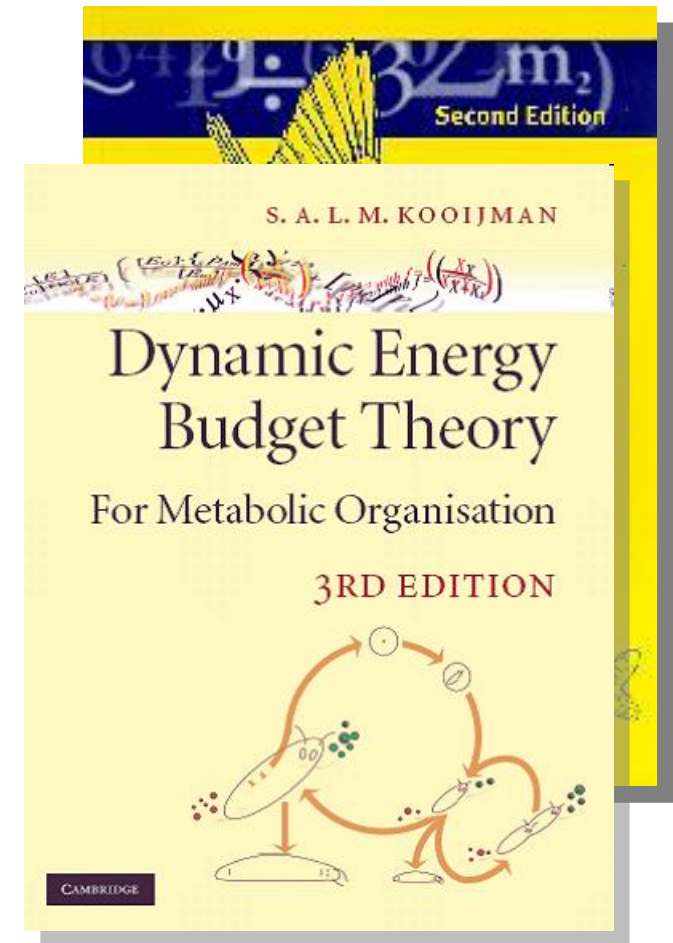
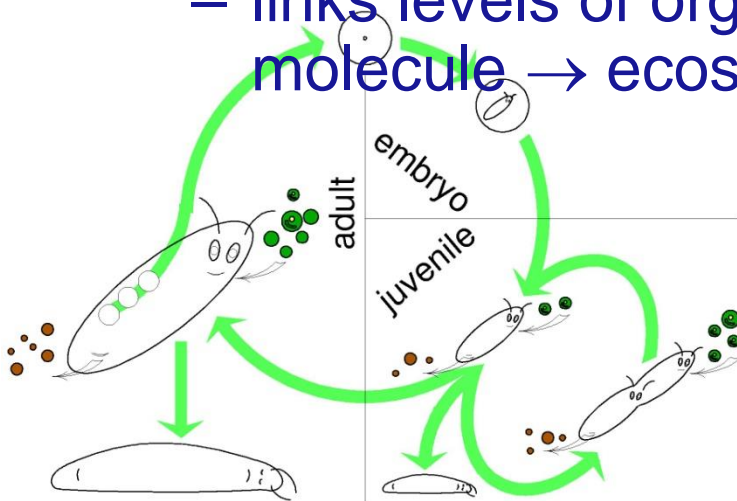
DEB MODELS

Quantitative theory for metabolic organisation from 'first principles'

- time, energy and mass balance

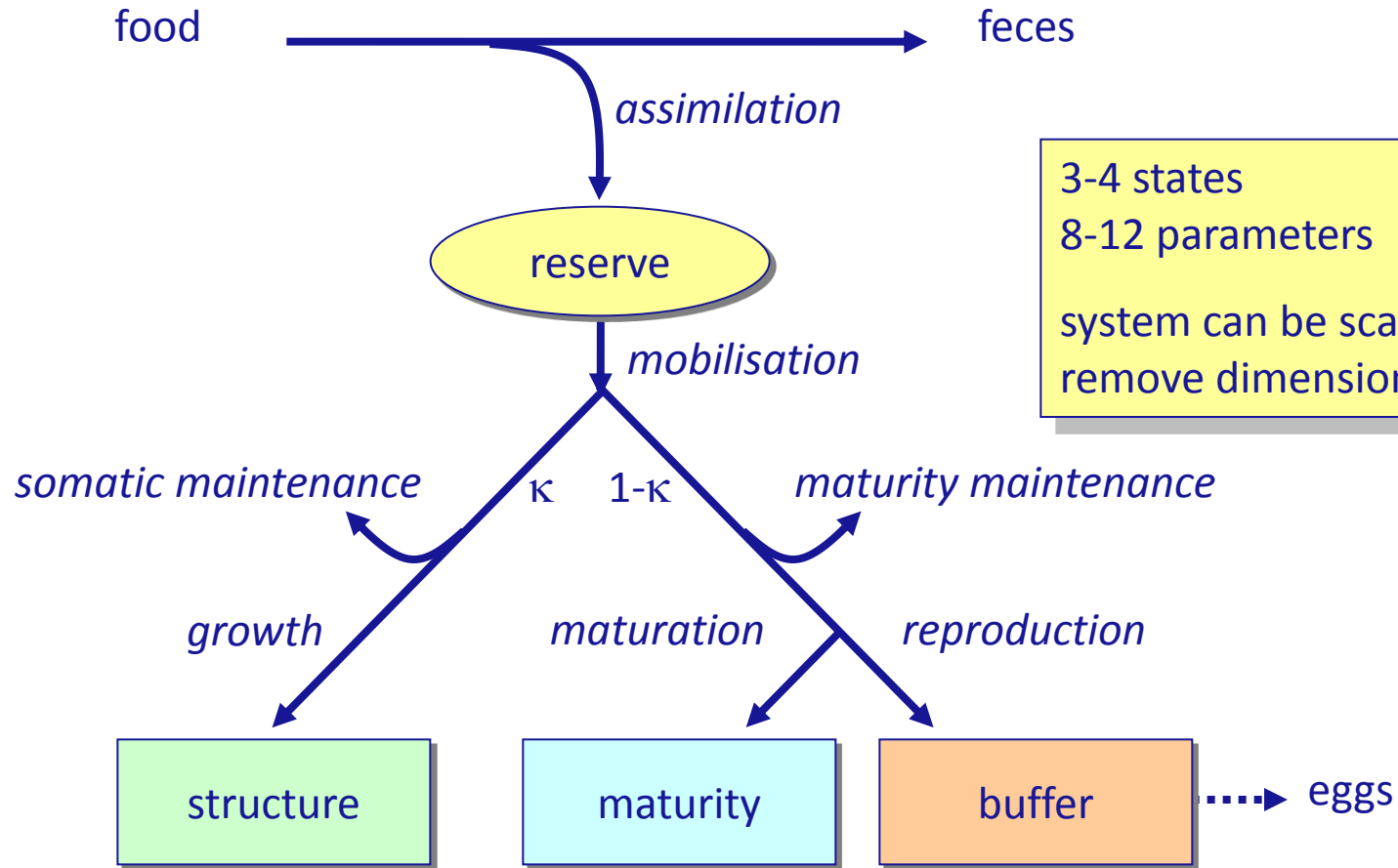
Life-cycle of the individual

- links levels of organisation: molecule → ecosystems

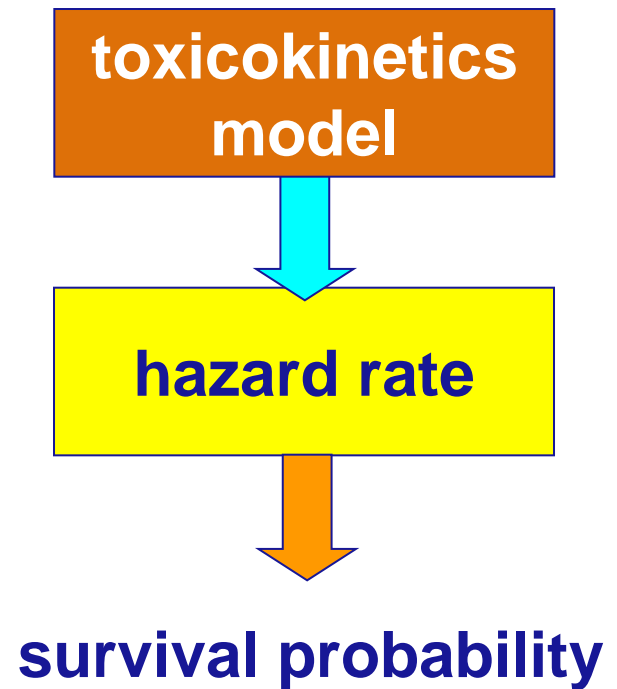
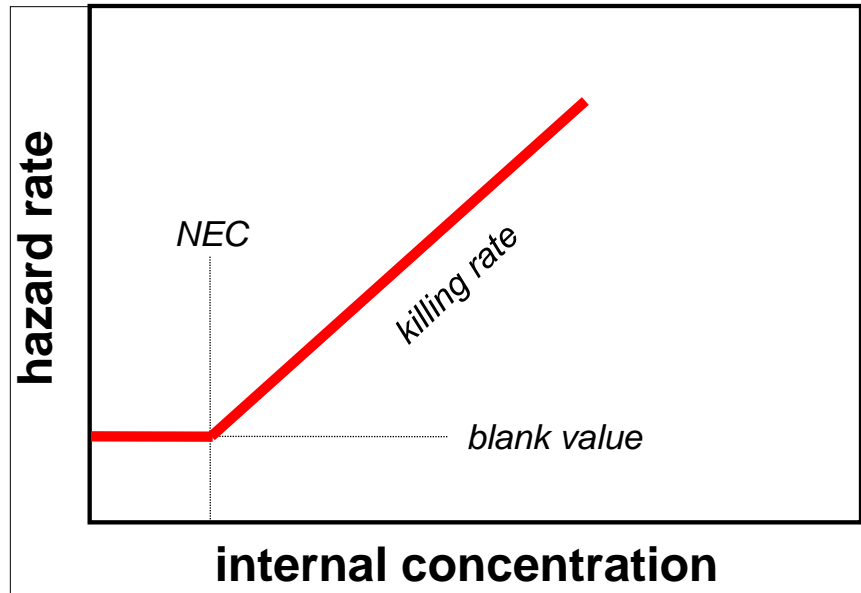


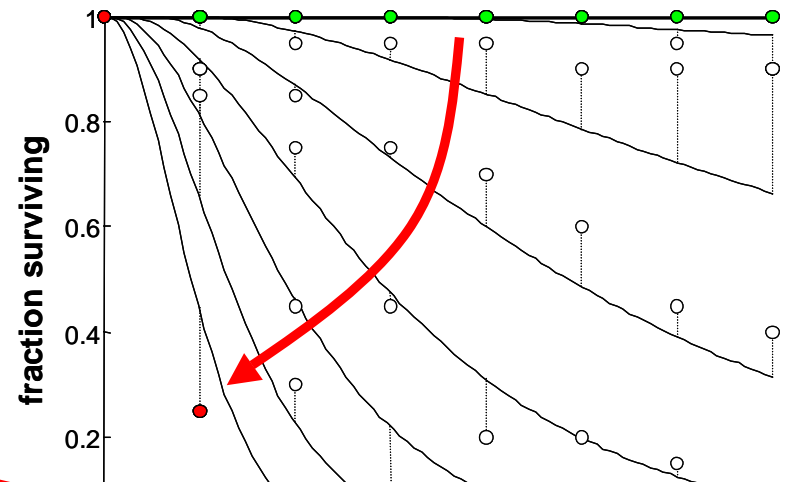
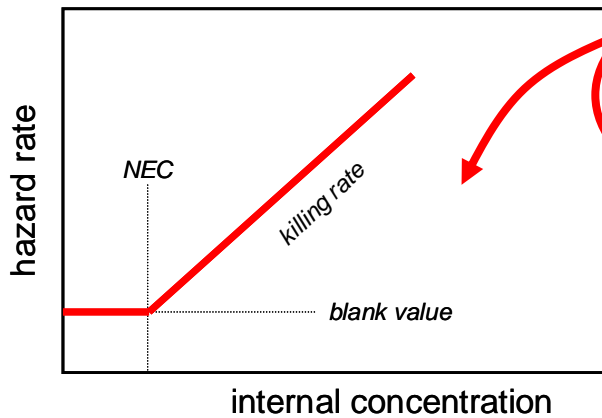
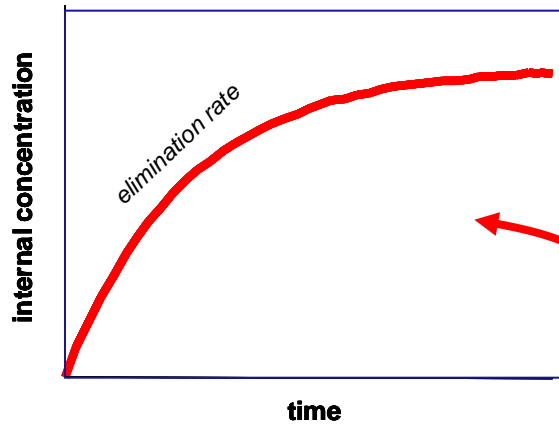
Kooijman (2010)

What are DEB MODELS ?

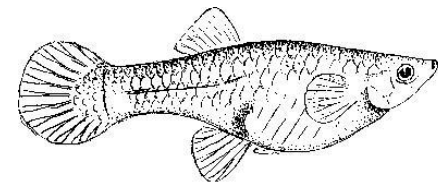


- Chemical affects the **probability** to die
 - hazard modelling

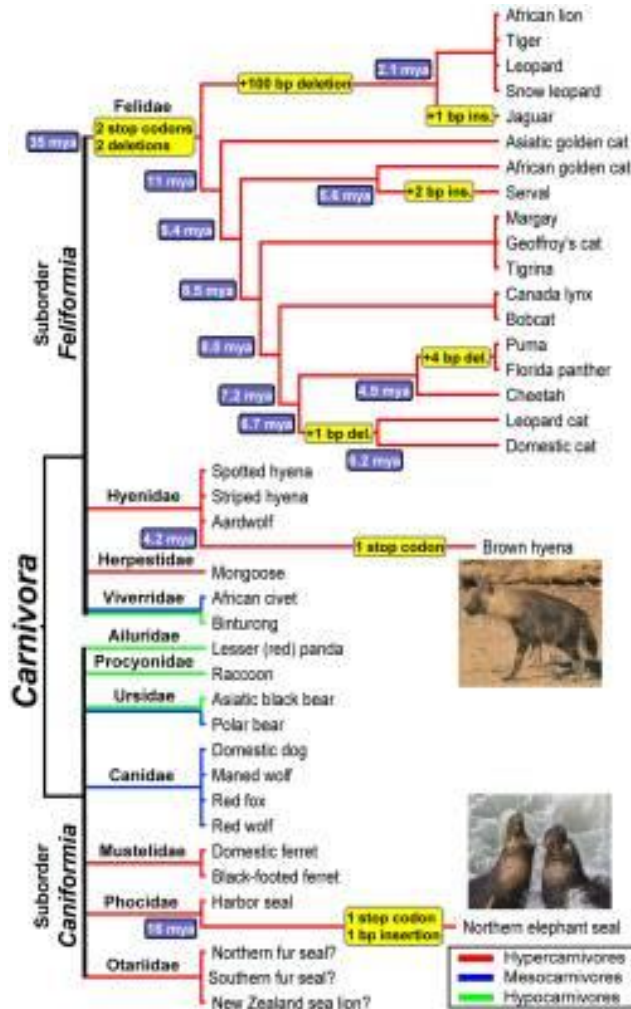




Elimination rate	0.73	d^{-1}
Blank hazard rate	0.0064	d^{-1}
NEC	2.8 (2.1-3.1)	$\mu g/L$
Killing rate	0.031	$L/(\mu g d)$



CAT EVOLUTION, HYPERCARNIVORY AND DIET



Cat in UGT enzymes (hypercarnivory)
no induction by plant compounds

Lacking

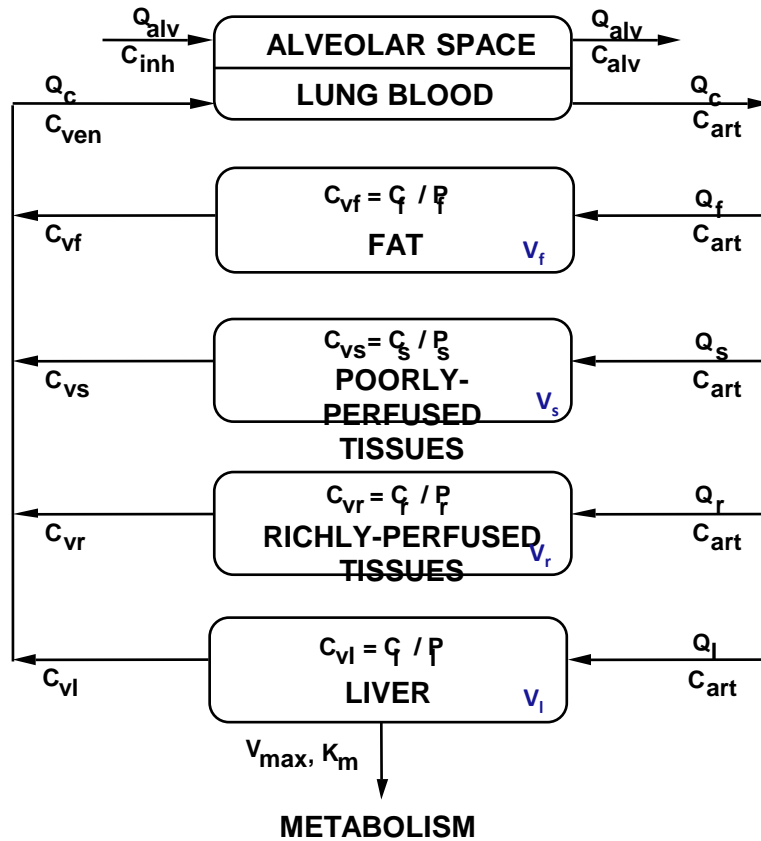
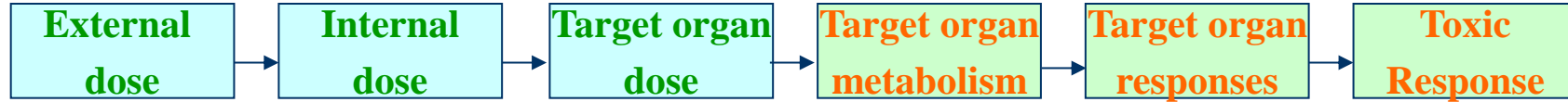
-Glycine conjugation

-N-acetyltransferases

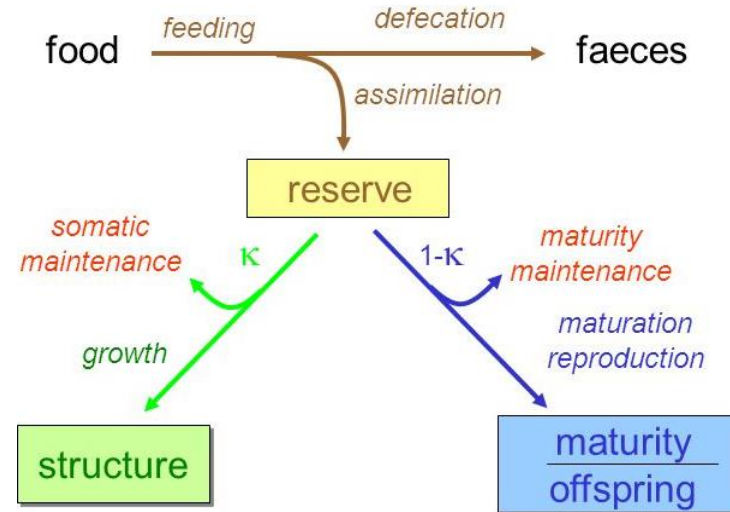
-Thiopurine s-methyltransferases

In context of Ecological RA and endangered species can we predict toxicity using physico-chemical properties, structure ?

-Building Open source TK and DEB tools-



PB-TK models



DEB Models



14 November 2016

Methodology

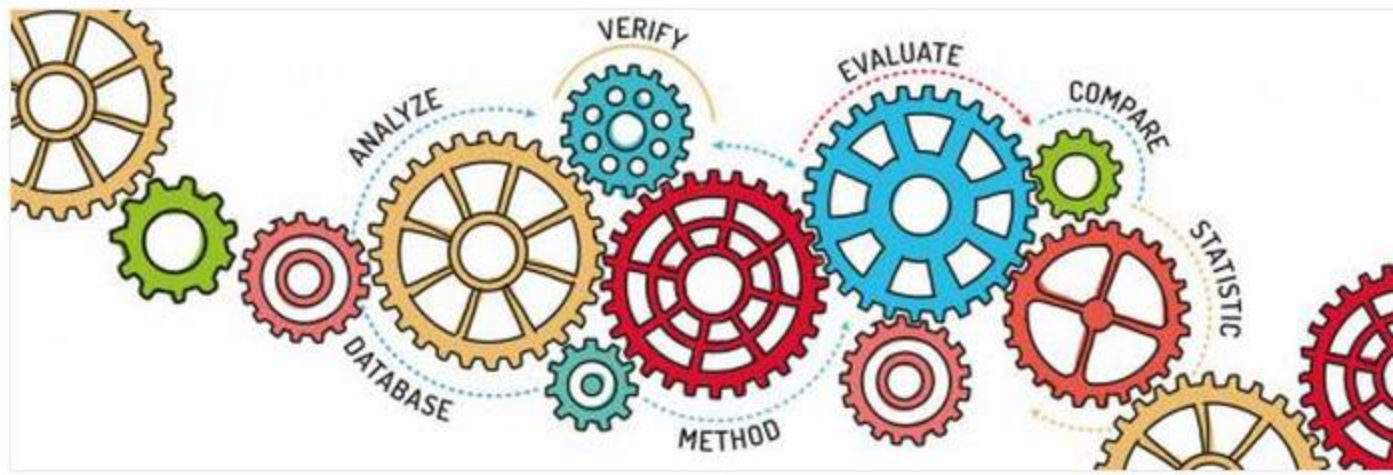
print

Tweet

Share

Share

Knowledge Junction – open access to scientific models



Scientific models used by EFSA over the last 15 years have been brought together in a new EFSA community: the **Knowledge Junction**. The models can be shared and cited and you can submit your own. A selection of these tools are also available as web applications on the new EFSA Statistical Models Platform; just

CONCLUSION AND RECOMMENDATIONS

- ✓ **OpenFoodTox** provide **historical data** from EFSA RA Human, animal health and Ecological RA
- ✓ **QSAR models** developed from OpenFoodTox **support 3Rs**
- ✓ **Open Source TK models** to integrate exposure (external) to TK (internal) and Toxicity **by 2018**
- ✓ **Open Source DEB models:** Taxa-specific data to link internal dose to toxicity **by 2018**
- ✓ **TK/TD platform:** Integration of population variability in TK and TD processes for RA **(by 2020-2021)**

THANKS TO PARTNERS

- ✓ **OpenFoodTox:** S-IN : Vicenza, Italy
- ✓ **QSAR models :** Istituto Mario Negri, Milan, Italy
- ✓ **TK platform:**
 1. Laser Analytica, Paris, France
 2. INERIS, Paris, France
 3. Radboud University, Nijmegen, The Netherlands

- ✓ **TK/TD platform:**
 1. ANSES, Paris, France
 2. ISS, Rome, Italy
 3. University of Utrecht, Utrecht The Netherlands
 4. University of Bretagne, Brest, France

- ✓ **DEB Models**
 1. Center for Ecology and Hydrology , UK
 2. Technico Lisboa, Lisbon, Portugal
 3. Akvaplan Niva, Tromsø, Norway

**Many Thanks
Questions ?**

