

# CLiCC (Chemical Life Cycle Collaborative)

## Network for Rapid Assessment of Chemical Life Cycle Impact



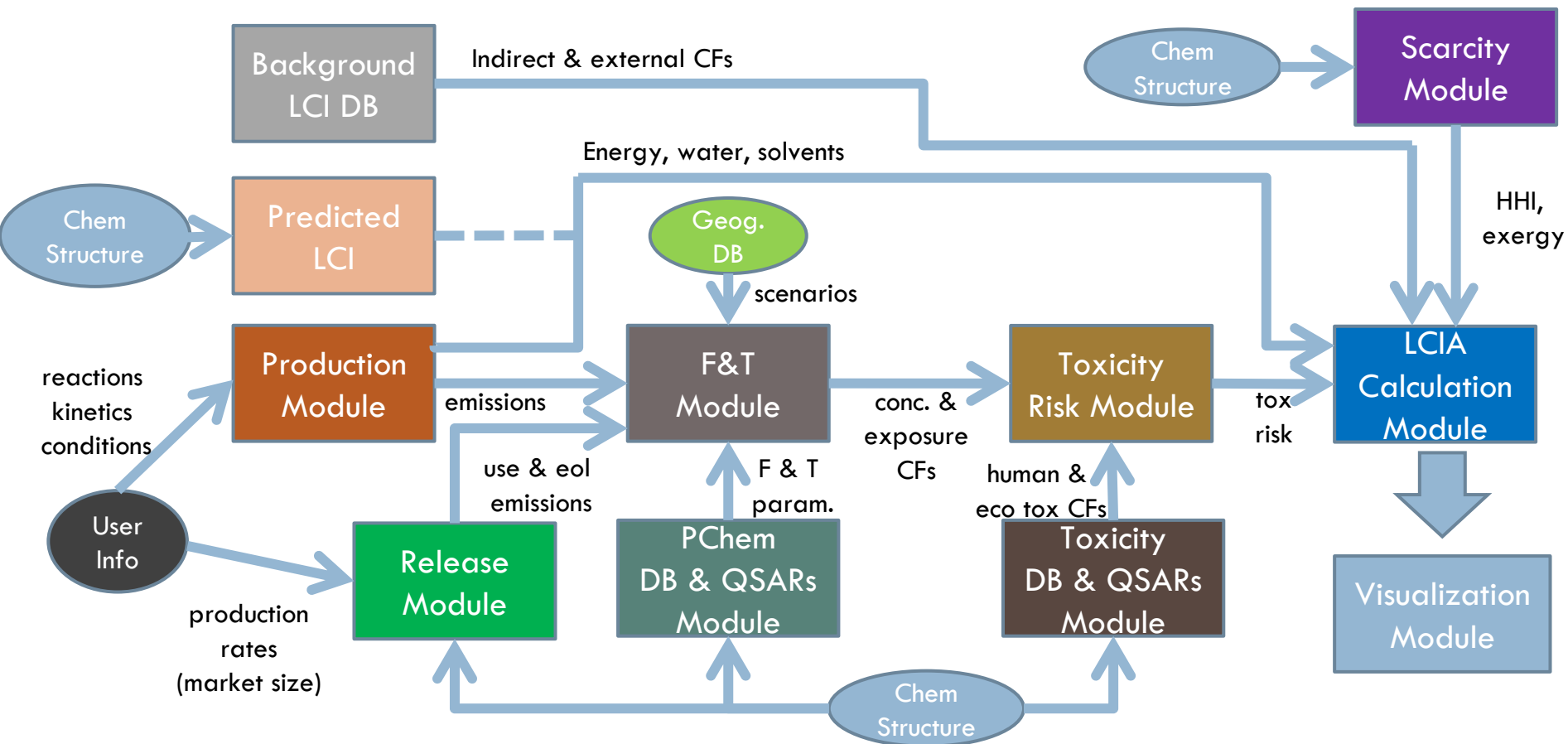
UCSB

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# Internal Tool Workflow

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# CHEMICAL LIFE CYCLE COLLABORATIVE: CHEMICAL PROPERTIES MODULE

Mengya Tao, Runsheng Song, Jaye Harada, Kristin Denault

# Chemical Properties Module

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- Extending search of QSARs to more endpoints:
  - Carcinogenicity (more quantitative)
  - Developmental toxicity
  - Reproductive toxicity
  - Cardiovascular toxicity
  - Dermatotoxicity
  - Endocrine toxicity
  - Epigenetic toxicity
  - Genotoxicity
  - Hematotoxicity
  - Hepatotoxicity
  - Immunotoxicity
  - Musculoskeletal toxicity
  - Neurodevelopmental toxicity
  - Neurotoxicity
  - Ocular toxicity
  - Respiratory toxicity
  - Skin sensitization

# Scarcity & Exergy

Evaluating proxies for natural resource and economic considerations

Jaye Harada

# Scarcity module: abiotic depletion methods applied to new inorganics

- Current abiotic resource depletion methods do not evaluate all aspects of scarcity
  - ▣ supply risk
  - ▣ depletion rate
  - ▣ ore grade decrease
- Three goals for module:
  - ▣ Calculation of existing abiotic resource depletion characterization factors for new inorganic materials
    - **User can choose which method(s) to use** to evaluate a material's scarcity
  - ▣ Integration of future resource demand and production scenarios to calculate **future-oriented characterization factors**
    - These factors will be limited to reserve-based and production-based methods
  - ▣ **Evaluate uncertainty** in USGS production and reserves data
    - Consider year-to-year variance in USGS assessments

# Potential exposure models at different levels

Dr. Dingsheng Li (new CLiCC project member)

Past work and ideas for the future

# Introduction

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## □ Why do we need exposure models



Environmental  
fate of  
chemicals

Human  
health  
impact  
assessment



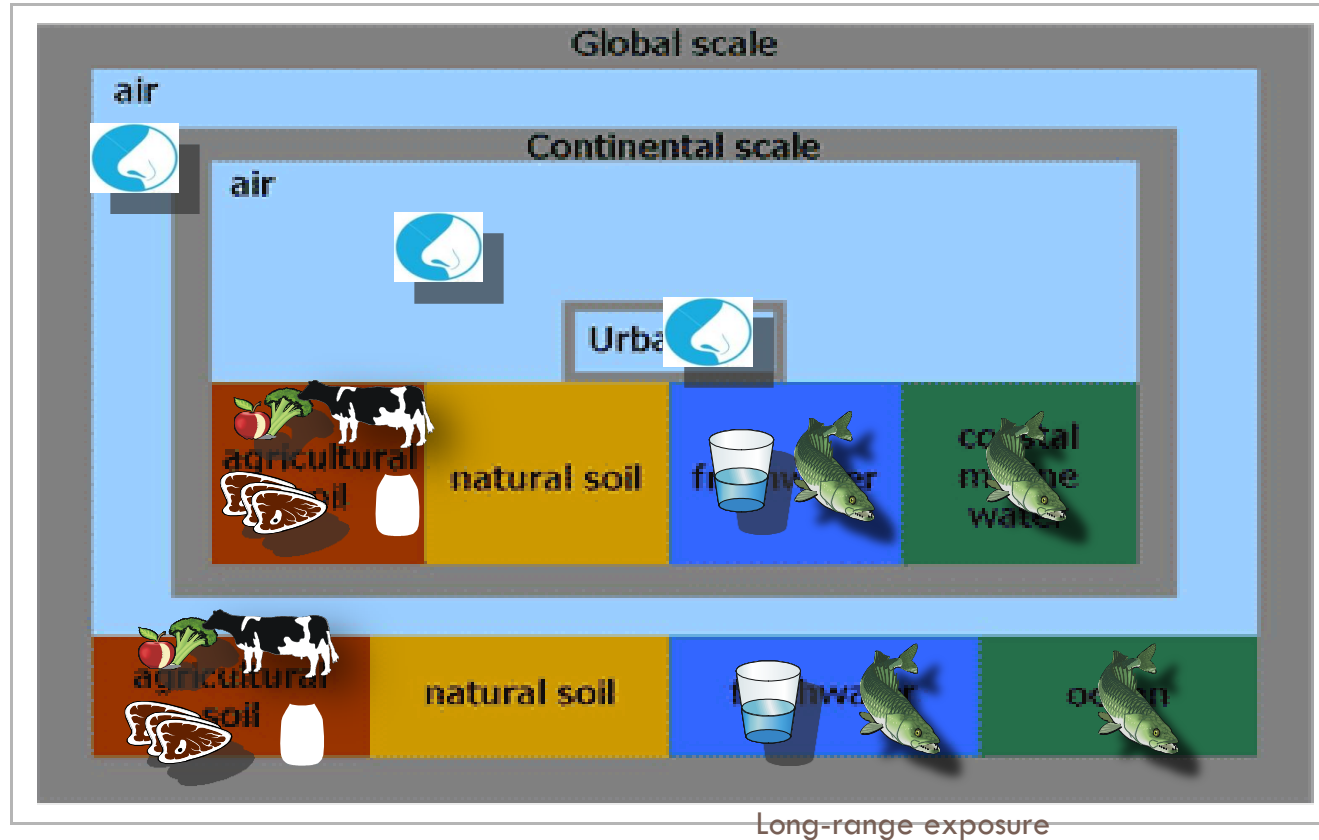
# Models at different levels

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- **Long-range exposure models**
  - ▣ Traditionally employed in life cycle impact assessment (LCIA)
  - ▣ Can be improved for specific categories of chemicals
- **Close-range exposure models**
  - ▣ Indoor exposure
  - ▣ Personal care products
- **Internal organ specific exposure model**
  - ▣ Potential use of physiologically based toxicokinetic (PBTK) model
  - ▣ Linking target organs with toxic effects of chemicals
  - ▣ Much more complex than the other two

# Long-range exposure

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# Input parameters

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## □ **Inhalation:**

- Concentration of chemical in the air (from Fate & Transport module)
- Inhalation rate of the population (from EPA exposure handbook, can be adjusted for sensitive population)
- Population size (pre-defined, scenarios)

## □ **Ingestion from water:**

- Concentration of chemical in the water (from Fate & Transport module)
- Ingestion rate of water (from EPA exposure handbook, can be adjusted for sensitive population)
- Population size (pre-defined, scenarios)

## □ **Ingestion from food:**

- Concentration of chemical in the water and agricultural soil (from Fate & Transport module)
- Bioconcentration factors, biotransfer factors, etc. (from previous empirical models, need support from QSAR for calculations)
- Ingestion rate of different produces (from established databases, can be differentiated to different age groups)
- Population size (pre-defined, scenarios)

# Output parameters

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## □ Intake amount

- ▣ Expressed in mass (kg) or dose (mg/kg-day)
- ▣ Can be converted to intake fractions ( $\text{kg}_{\text{intake}}/\text{kg}_{\text{emitted}}$ )
- ▣ Used to estimate human toxicity impact
- ▣ Requires either epidemiology data or chronic *in vivo* animal toxicity data, which can be supported by QSAR module

# Most suitable for

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- **Chemicals emitted to the general environment**
  - ▣ Byproducts, pollutants, pesticides, etc.
  - ▣ No need to address indoor exposure/dermal exposure
- **For chemicals with relatively limited physico-chemical data**
  - ▣ Missing data can be generated from the QSAR module or Fate & Transport module

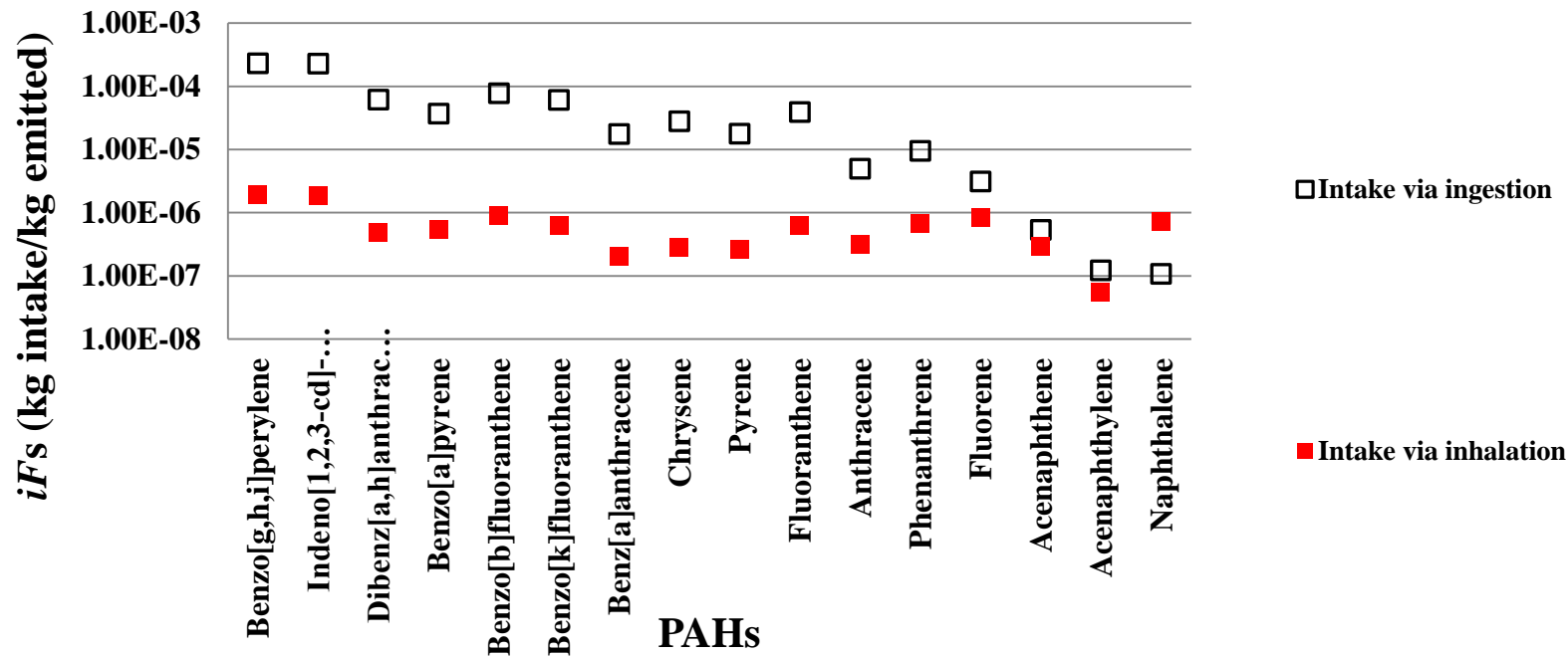


Long-range exposure



# Example

14



Long-range exposure

# Close-range exposure

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FAN1006953 [RF] © www.visualphotos.com

Close-range exposure

# Input parameters

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## □ **Inhalation:**

- ▣ Removal and degradation rates (from indoor air model [Wenger et al., 2012])
- ▣ Inhalation rate of the population (from EPA exposure handbook, can be differentiated to different age groups)
- ▣ Indoor room descriptions: ventilation rate, volume, occupants, temperature, etc. (pre-defined, scenarios)

## □ **Dermal exposure:**

- ▣ Contact duration (pre-defined, scenarios, data from industry)
- ▣ A series of permeability and transfer rates (QSAR, Berg 2009)

Close-range exposure



# Output parameters

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## □ Intake fractions

- ▣ Expressed in fractions ( $\text{kg}_{\text{intake}}/\text{kg}_{\text{emitted}}$ )
- ▣ Usually orders of magnitude higher than iF of the same chemicals released to the general environment
- ▣ Used to estimate human toxicity impact, with support of emitted/applied mass (user input)
- ▣ Requires either epidemiology data or chronic *in vivo* animal toxicity data, which can be supported by QSAR module

# Most suitable for

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- **Chemicals emitted to the indoor environment**
  - ▣ VOCs that are released from products used indoors
  - ▣ Occupational setting
- **Chemicals used in personal care products**
  - ▣ Directly applied to skins such as shampoo, lipsticks, lotions, etc.
  - ▣ Data about how the products are used is essential

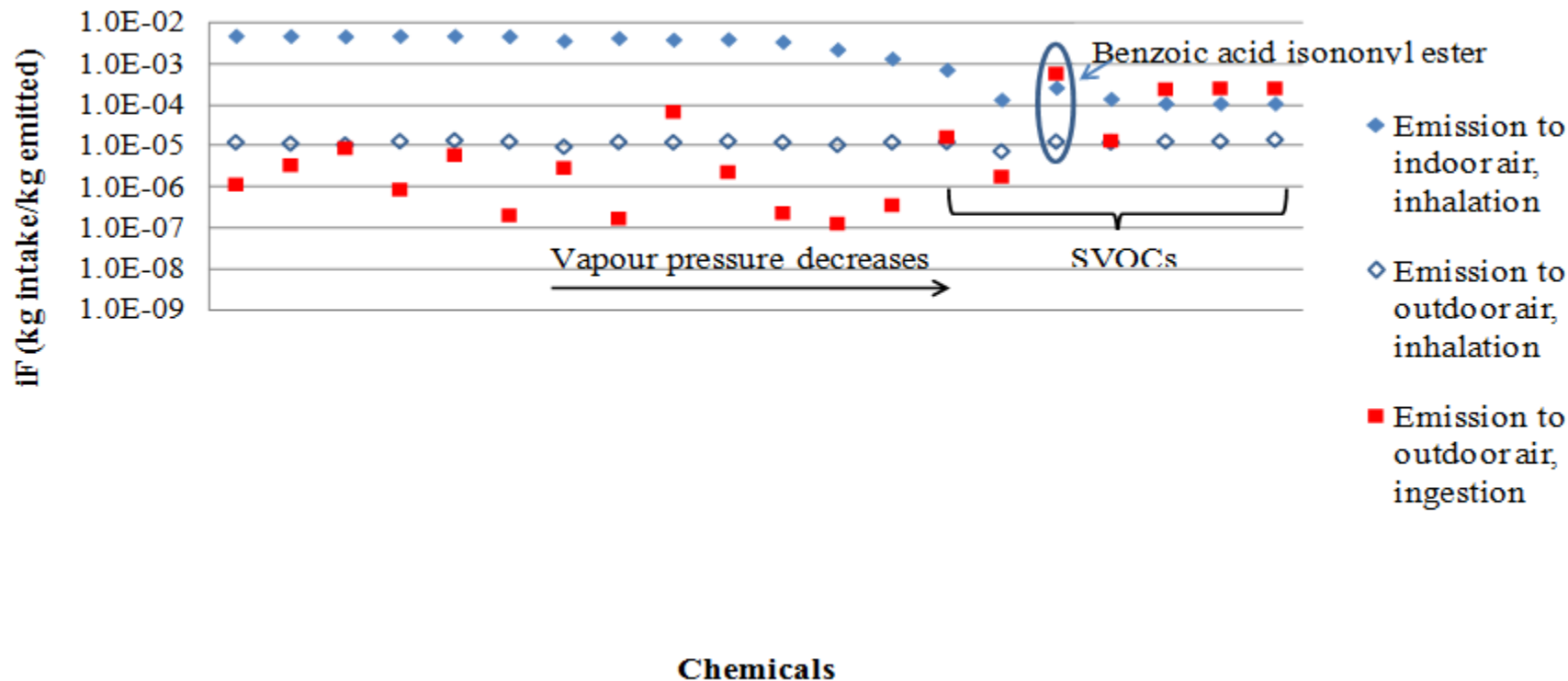


Close-range exposure



# Example

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Close-range exposure

# Example

20

4 min

8 hrs

Time [hr]

0.0001

0.001

0.01

0.1

1

1

10

$iF$

0.1

0.01

0.001

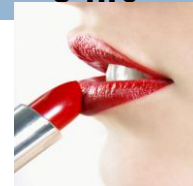
0.0001

0.00001

high volatility, high skin permeability

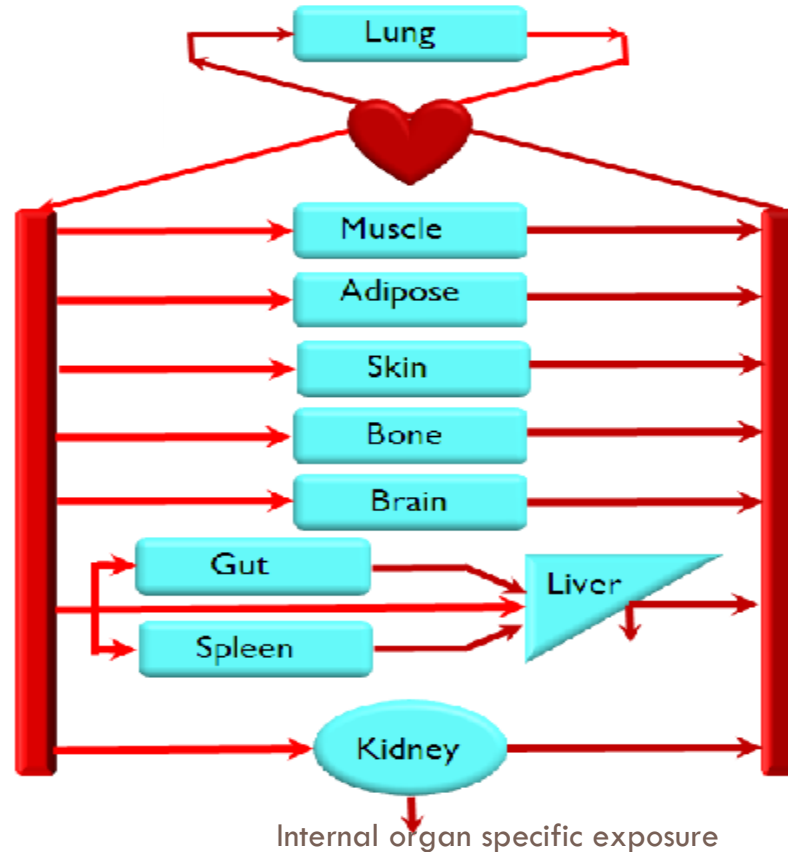
low volatility, low skin permeability

Close-range exposure



# Internal organ specific exposure

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# Input parameters

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- **Human physiology data:**
  - ▣ Body weight, organ weights, cardiac output, etc.
  - ▣ Existing literature, can be adjusted for sensitive population
- **Exposed amounts:**
  - ▣ Concentration of chemical in air/food (from Fate & Transport module)
  - ▣ Inhalation rate and ingestion rates (from USEtox refs, EPA exposure handbook)
- **Inside body kinetics (most challenging):**
  - ▣ Adsorption (from existing QSAR type models: Caco-2, PAMPA, etc.)
  - ▣ Distribution (from existing database, potential QSAR models)
  - ▣ Metabolism (from existing database)
  - ▣ Excretion (from existing models)

# Output parameters

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- **Concentrations in blood and various organs**
  - ▣ Can be converted to cumulative amounts in blood and various organs over time
  - ▣ Compare with high throughput *in vitro* toxicity tests
  - ▣ Can be independent on epidemiology/animal tests, opening up much wider toxicity dataset
  - ▣ More accurate representation of internal dose – given the data and model are good (otherwise, garbage in garbage out)

# Most suitable for

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- **Chemicals that require higher accuracy or dynamic of exposure**
  - ▣ Can predict doses in sensitive organs at different ages
- **Chemicals without epidemiology/animal toxicity data**
  - ▣ Can use other data sources for human health impact assessment
- **Chemicals with richer physiological kinetic data**
  - ▣ Linked with QSAR models, the data gap in the ADME parameters may be closed

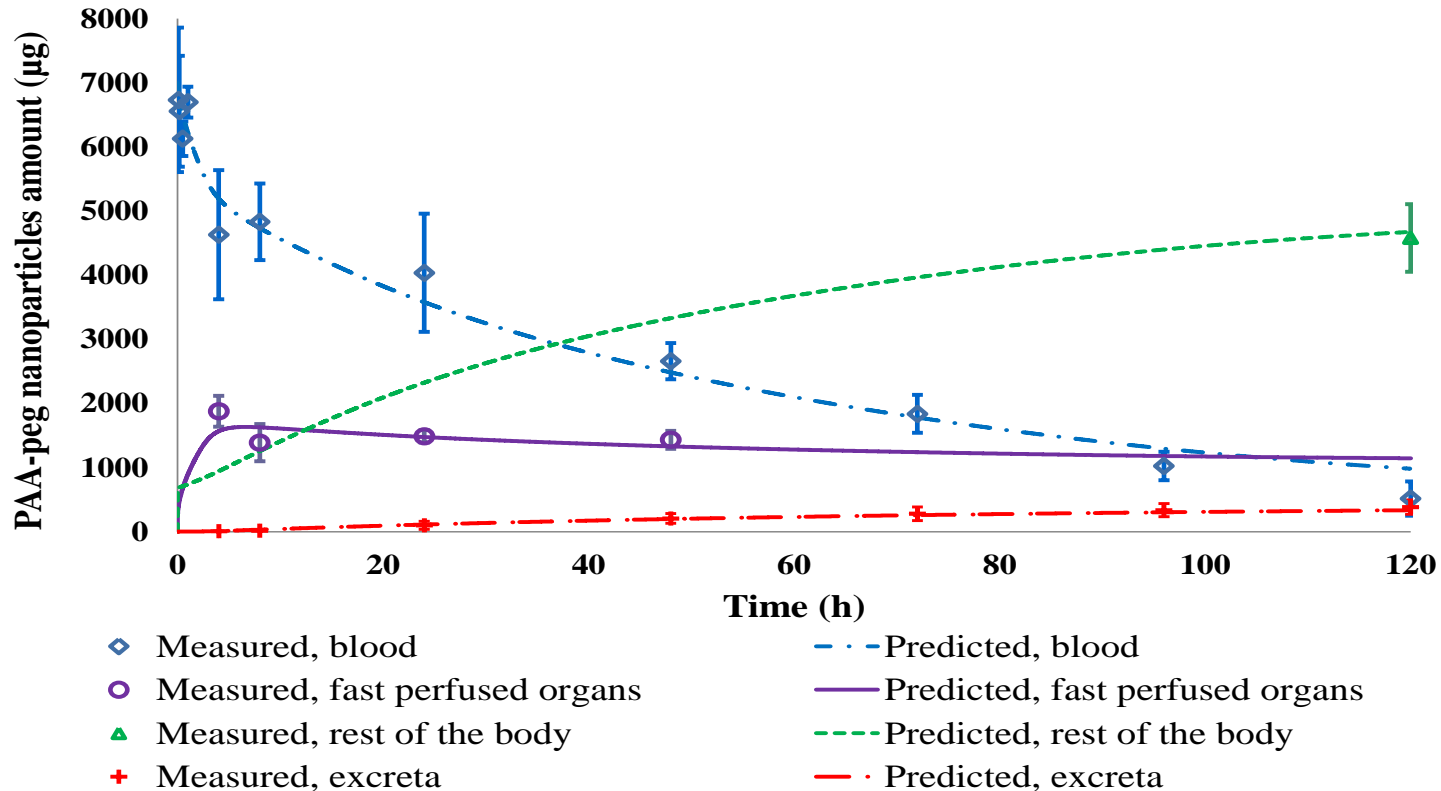


Internal organ specific exposure



# Example

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Internal organ specific exposure

# More of a field of research

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- **No “generic” PBTK model exist yet**
  - The community in toxicology is still working on this topic
  - Mostly due to the complexity of different chemicals kinetics inside the body
- **No attempt to link PBTK with LCIA has been made**
  - Cross disciplinary may LCIA be, PBTK is still untouched by LCIA researchers
- **More complex model, more computation time**
  - Even the most basic PBTK model is much more complex than the other exposure models
  - Complexity similar to multimedia environmental fate models
  - Therefore takes more computation power
  - More complexity usually leads to more uncertainty, too

Internal organ specific exposure

# Summary

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- Three models addressing different scales of exposure/output are being considered for further development and integration into CLiCC framework
- User would have the option to determine which one to use based on their need
- Ranking of readiness:
  - ▣ Long-range exposure models (easy after F&T model fully developed)
  - ▣ Close-range exposure models (relatively easy after F&T model fully developed)
  - ▣ Internal organ specific models (requires more complex PBTK modeling & QSARs for internal body parameters)

# APPLICATION OF THE CLiCC TOOL

USES FOR DECISION MAKERS

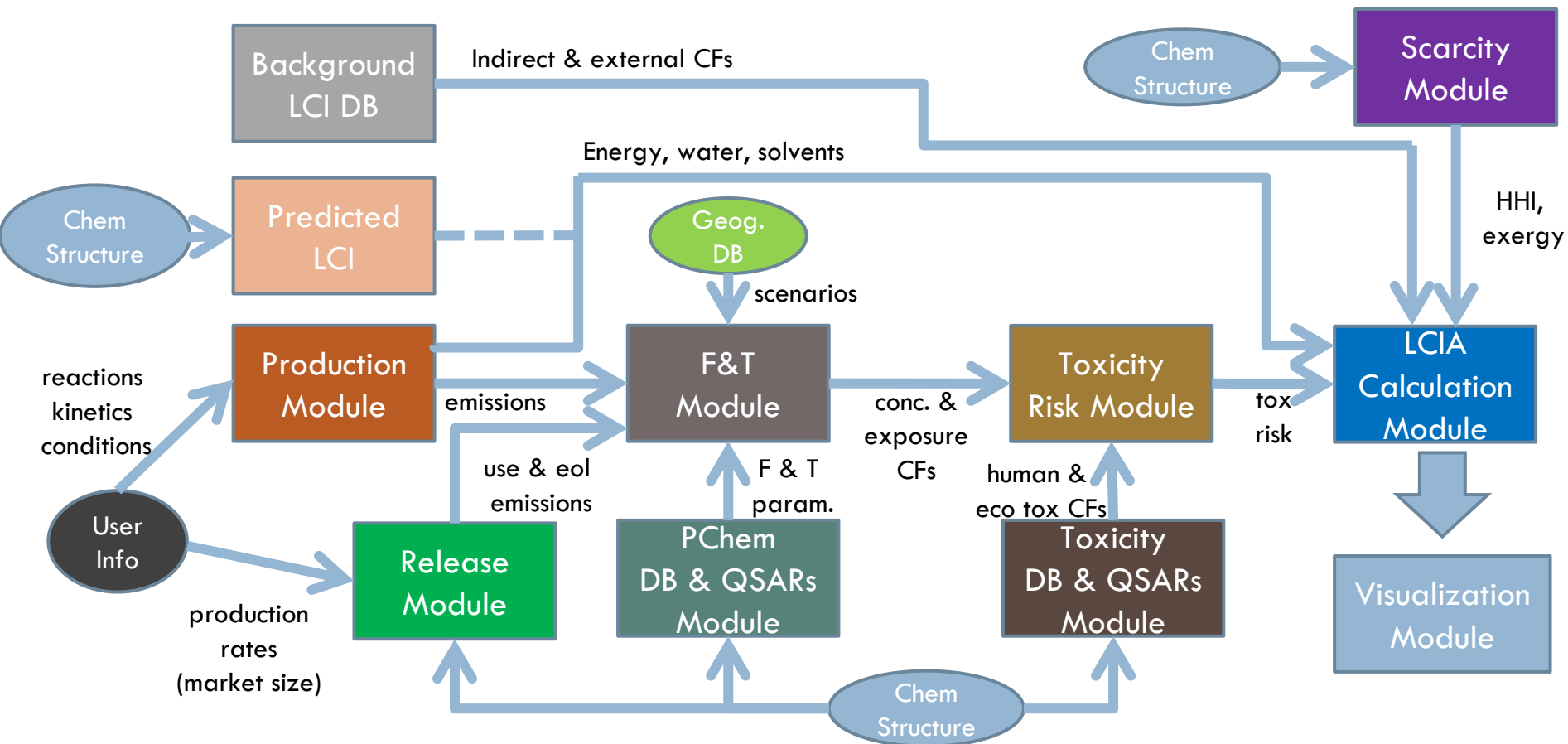
# Industry Partner Guidance

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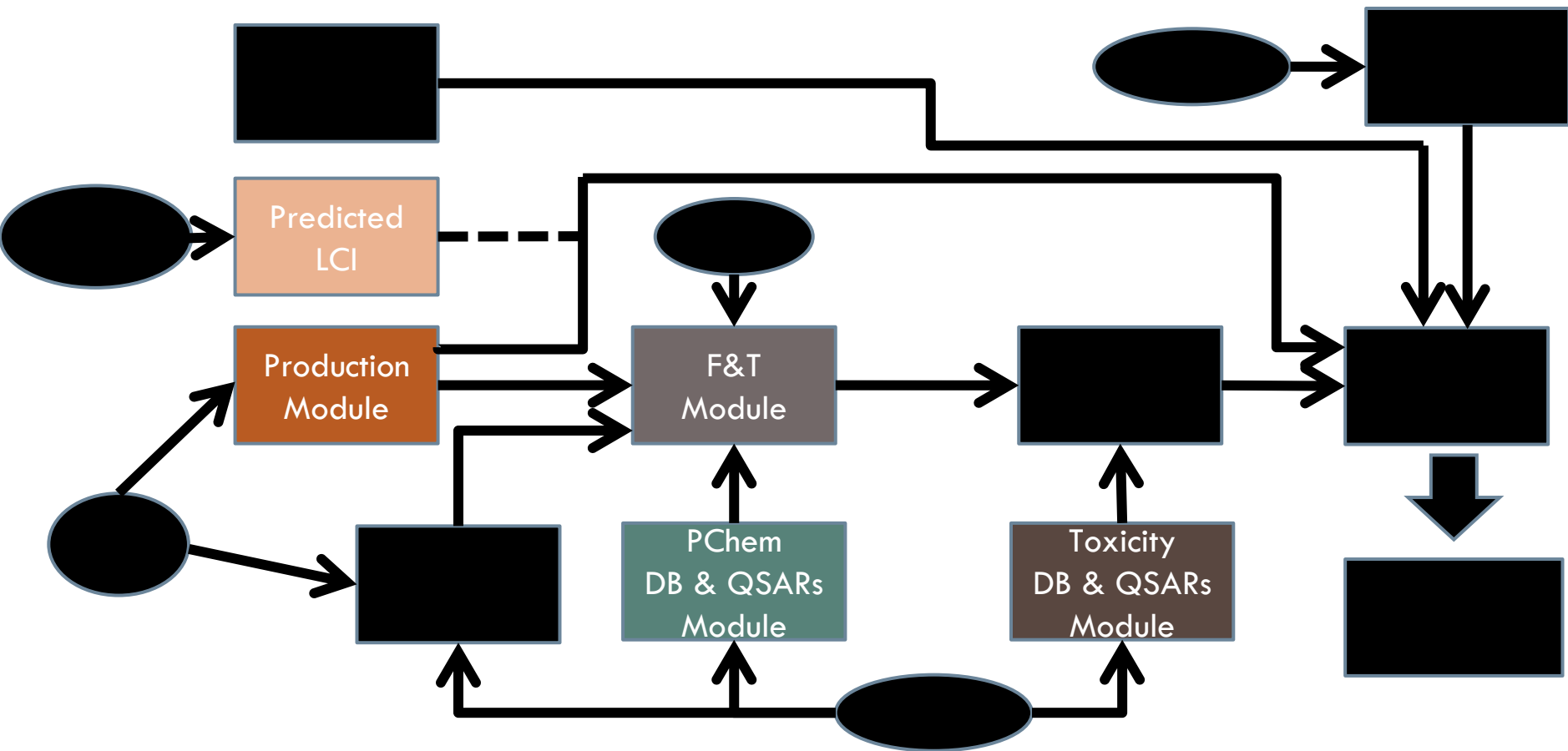
- Characterization of uncertainty is very important
  - ▣ Identification of uncertainty “hotspots”
  - ▣ Stochastic representation (probability distributions)
- Outputs provided for individual modules (not just entire CLiCC Tool results)
  - ▣ First round of case studies: individual modules to determine feasibility and guide output visualization
- Most users will be relatively technical and LCA literate
- Need to be transparent about data sources
  - ▣ Will provide output identifying the data sources used in each module for a given chemical run through the CLiCC tool

# CLiCC Tool Architecture

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# Modules for Case Studies





Reaction kinetics  
Chemical formula

Operating ranges  
Scale  
Upper/lower flammability limits  
Any other process constraints

Chemical  
Synthesis &  
Production  
Module

Raw material  
requirements (molar  
or mass)  
  
Production rates for  
product, by-product,  
waste  
  
Upper bound on  
selectivity  
  
Minimum energy requirements  
(next step in module development)

INPUTS



OUTPUTS



## SMILES

(simplified molecular  
input line entry system)

unique identifier for a given  
chemical that can be derived  
from its chemical structure,  
chemical name, or CAS number

## Predictive Life Cycle Inventory (LCI) Module

Artificial Neural  
Networks

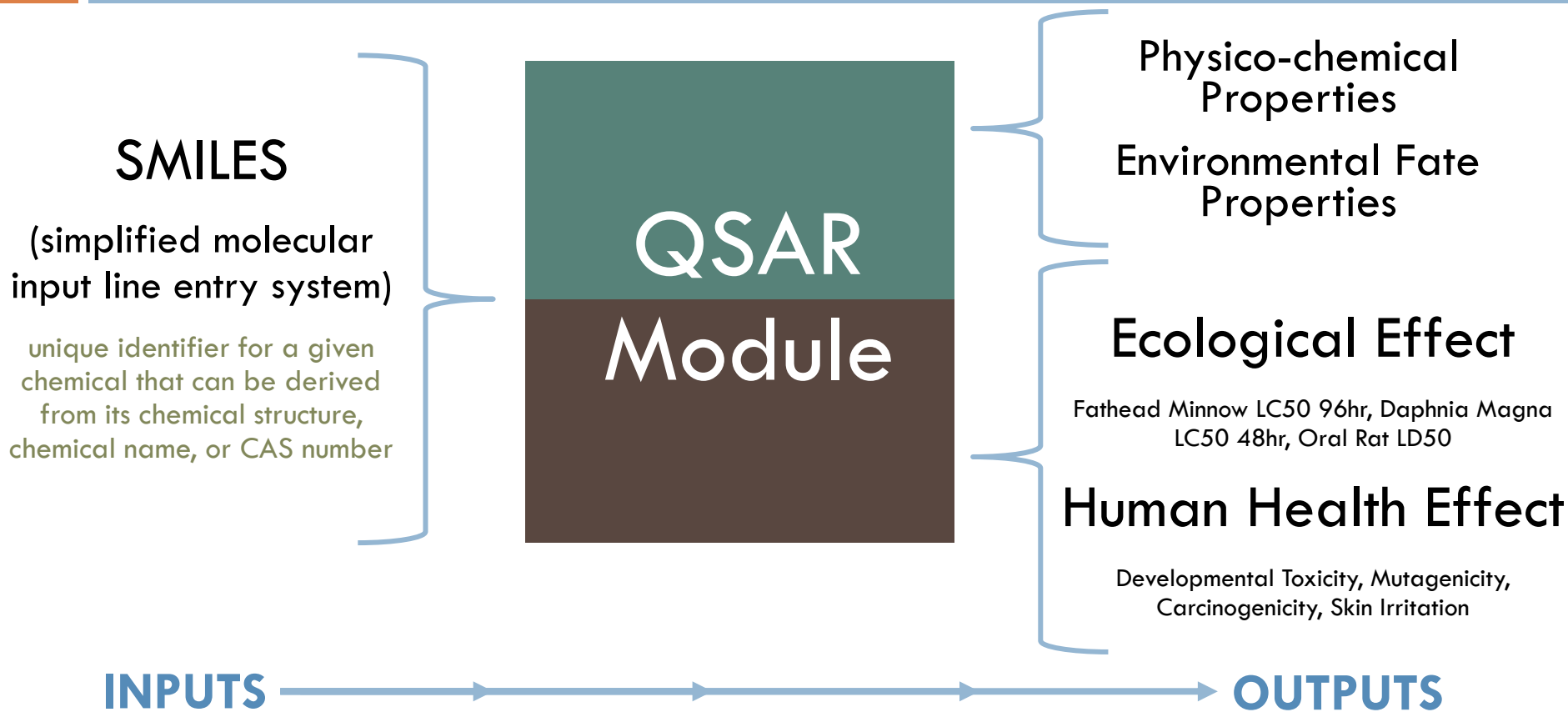
Cumulative Energy  
Demand (CED)

Water  
requirements

Global Warming  
Potential (GWP,  
CO<sub>2</sub> equivalents)

INPUTS

OUTPUTS



**Chemical  
name/identifier**

physico-chemical properties and  
environmental fate properties  
from QSAR module

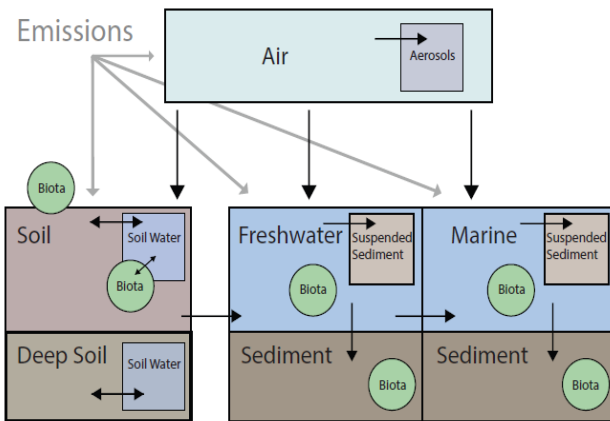
**Geographical Information**

**Emissions**  
(rate of release over time)

# Fate & Transport Module

## Fate Factors

(FF, concentration and mass by  
compartment over time)



**INPUTS**

**OUTPUTS**

Thank you!