

A Knowledge Graph for Ecotoxicological Risk Assessment and Effect Prediction

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Ecological Risk Assessment



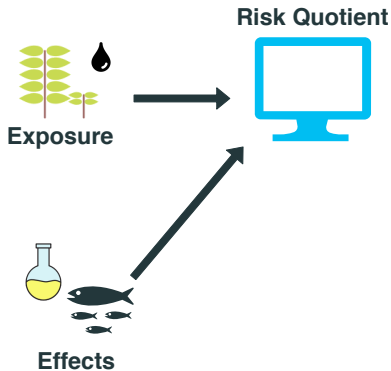
Risk assessment is an estimation of cumulative risk on individuals, populations, communities, and ecosystems from chemical pollutants.

Ecological Risk Assessment



Effect concentrations are found using organism experiments.

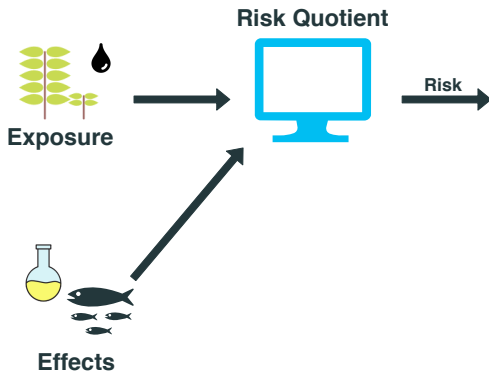
Ecological Risk Assessment



$$RQ = \frac{\text{environmental concentration}}{\text{effect concentration}}$$

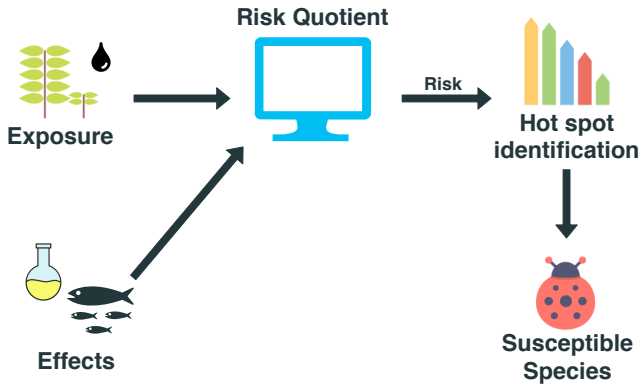
RQs coverage is limited by effect concentration experiments.

Ecological Risk Assessment



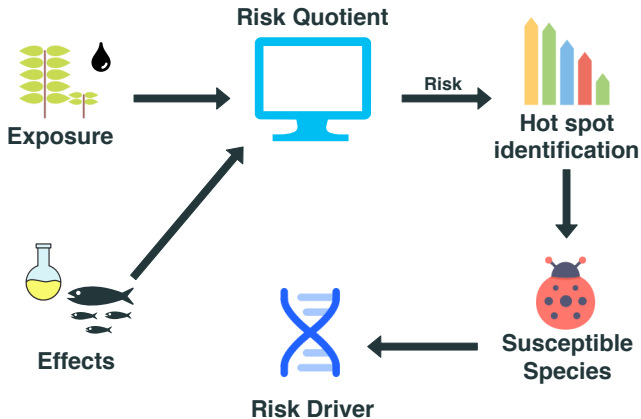
$$\text{risk}_{\text{group}} \approx \sum^{\text{chemicals}} RQ$$

Ecological Risk Assessment



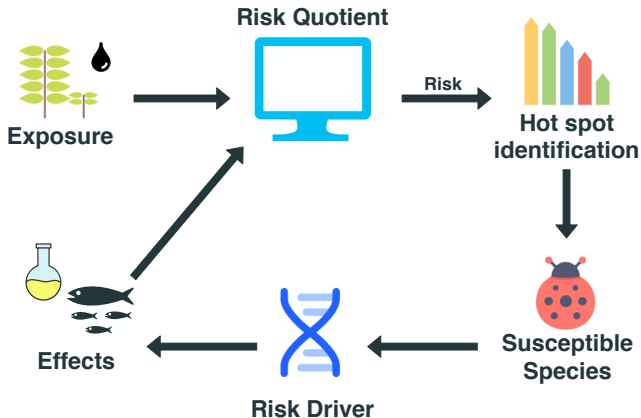
The risk is used to find further susceptible species.

Ecological Risk Assessment



Risk driver describes *how* the chemical affects an organism.

Ecological Risk Assessment



New effect hypothesis are then tested in the laboratory.

Project Goals

1. Construct a knowledge graph that integrates the relevant knowledge for ecological risk assessment.

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 - 2.2. Predict effect concentrations (regression).

Research Questions

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- RQ1. Can the disparate data sources used in ecotoxicological risk assessment be integrated into a knowledge graph to improve accessibility?
- RQ2. Can the knowledge graph be used to improve (or diversify) ecotoxicological effect prediction over current state-of-the-art models?

The TERA Knowledge Graph

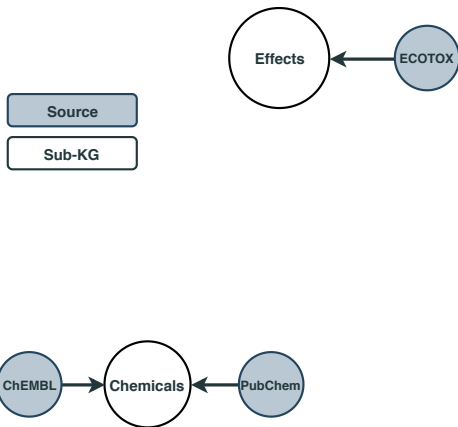
The Toxicological and Risk Assessment (TERA) knowledge graph integrates data sources varying in format.

The TERA Knowledge Graph



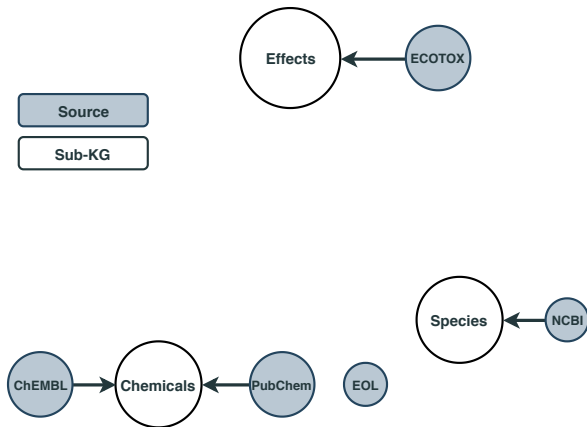
ECOTOX is the largest (public) source of effect data.

The TERA Knowledge Graph



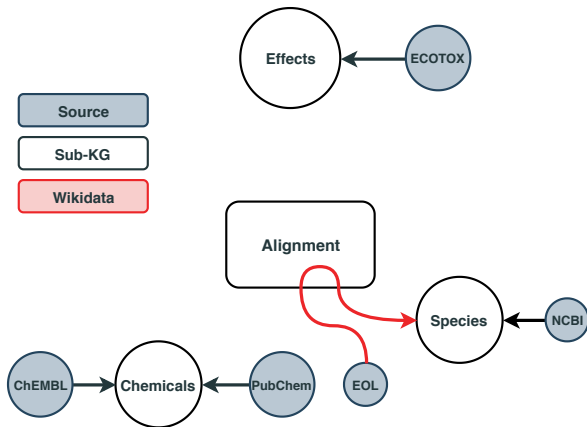
Importing the ChEMBL and PubChem knowledge graph.

The TERA Knowledge Graph



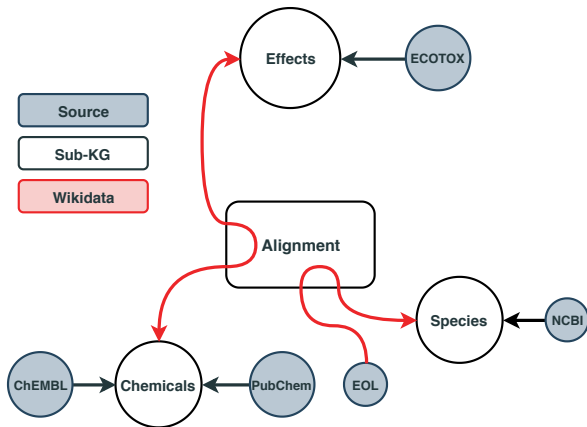
NCBI's tabular taxonomy and EOL traits are converted to a hierarchy.

The TERA Knowledge Graph



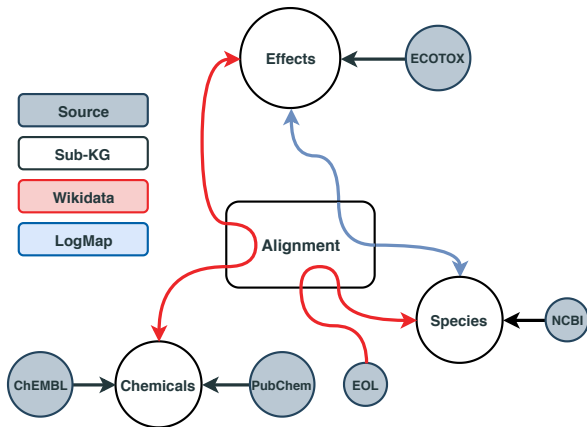
Encyclopedia of Life (EOL) is aligned to NCBI through Wikidata.

The TERA Knowledge Graph



Aligning proprietary chemical identifiers in ECOTOX to open identifiers in PubChem.

The TERA Knowledge Graph



Aligning taxonomies using ontology alignment tool LogMap.

Binary Effect Prediction Problem Definition

Binary Effect Prediction Problem Definition



C_1



C_2



C_3

Chemicals

Binary Effect Prediction Problem Definition



C_1



S_1



C_2



S_2



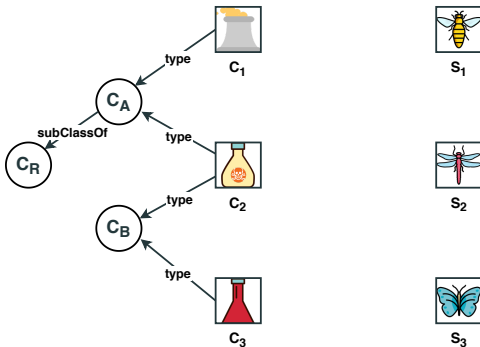
C_3



S_3

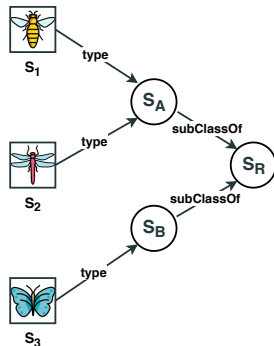
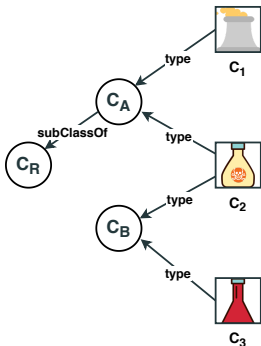
Species

Binary Effect Prediction Problem Definition



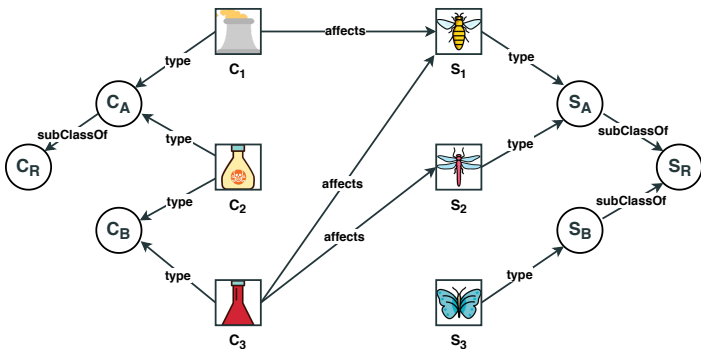
Chemical classification

Binary Effect Prediction Problem Definition



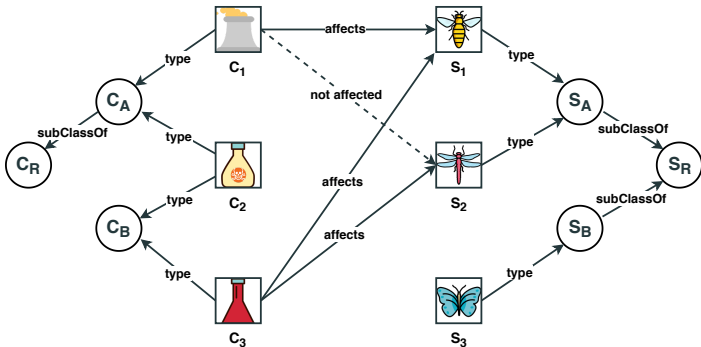
Taxonomy

Binary Effect Prediction Problem Definition



Positive samples

Binary Effect Prediction Problem Definition



Negative samples

Three models

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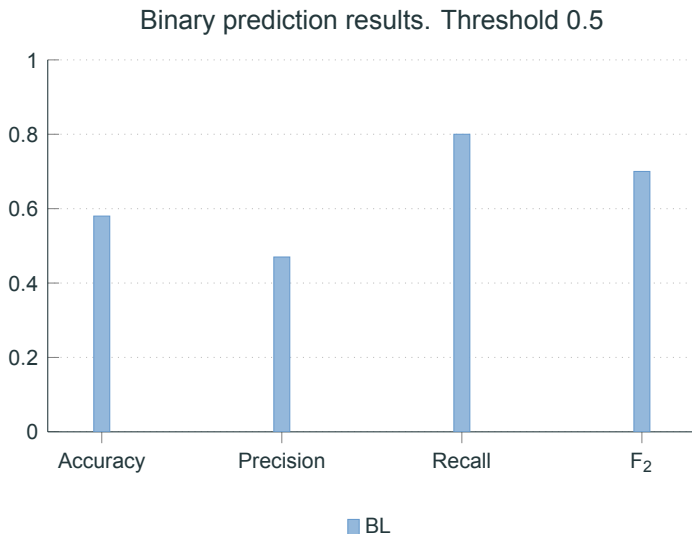
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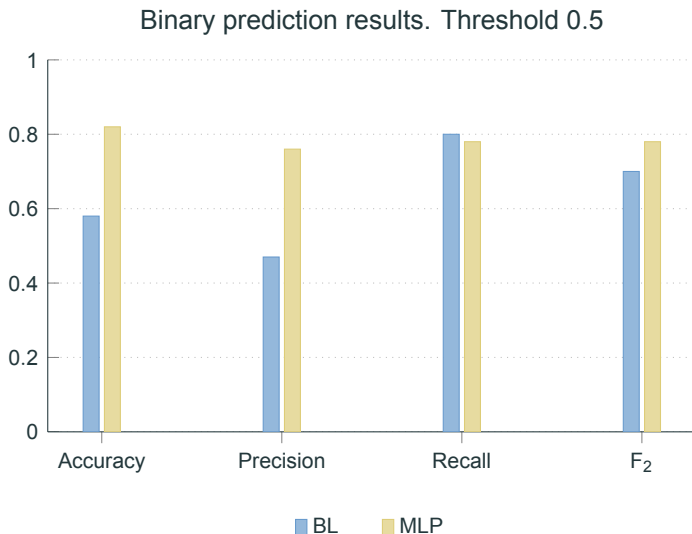
Knowledge graph embedding (KGE)

Replacing the input layers in MLP with a KGE model (e.g., TransE, DistMult, HolE).

Preliminary Results

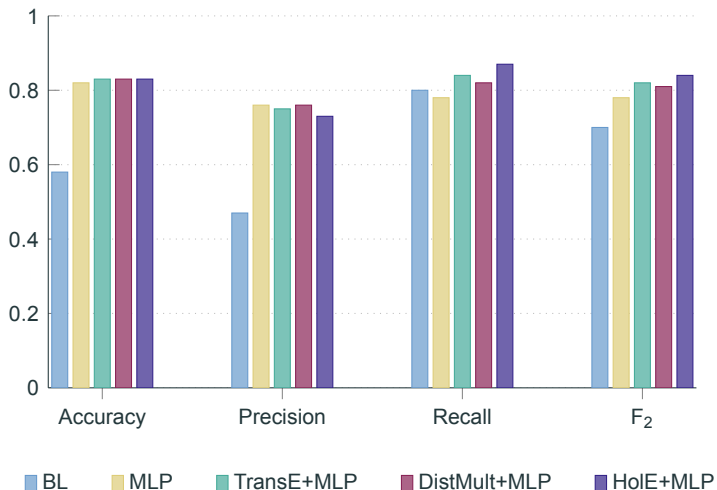


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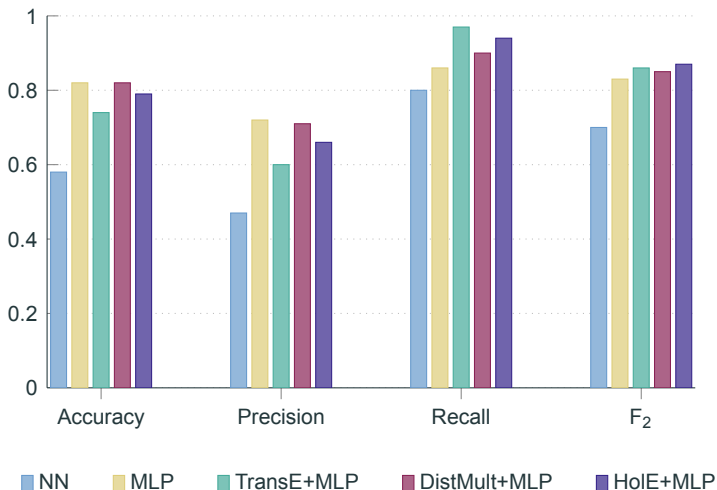
Preliminary Results

Binary prediction results. Threshold 0.5



Preliminary Results

Binary prediction results. Threshold 0.3



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RQ1 ☐ Integration of work into ecological risk assessment pipeline.

Interested in more details?

Come to the presentation tomorrow at 12.20 in Lecture Theater 098.



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