



TerraChem Pro

Leveraging Transfer Learning for Predicting Chemical Half-Lives in Mammals to Screen for Bioaccumulation in Air-Breathers

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Introduction

Elimination half-lives $elimt_{1/2}$ have been proposed as being more accurate than traditional $\log K_{OA}/K_{OW}$ based approaches to screen for bioaccumulation of chemicals in air-breathers^[1]. Still, the amount and accessibility of experimental data on $elimt_{1/2}$ is limited. Thus, **deep learning models**, like TerraChemPro, are promising non-animal **alternatives** to overcome these **data gaps** in future.

Objectives

Creating and validating a **novel screening tool** to identify chemicals as potentially (very) **bioaccumulative (vB/B)** in **mammals** to decide on regulatory follow-up assessment:

- robust screening criteria for big data
- as line of evidence for vB/B assessment
- open-access +easy to use for regulators

Ongoing works

Software update (v3.0)

- curation of toxicokinetic data (species/administration type/route)
- implementation of a batch mode
- comparisons with *in vitro* studies
- predicts $elimt_{1/2}$ dose depended (single vs. repeated doses)

Methodology

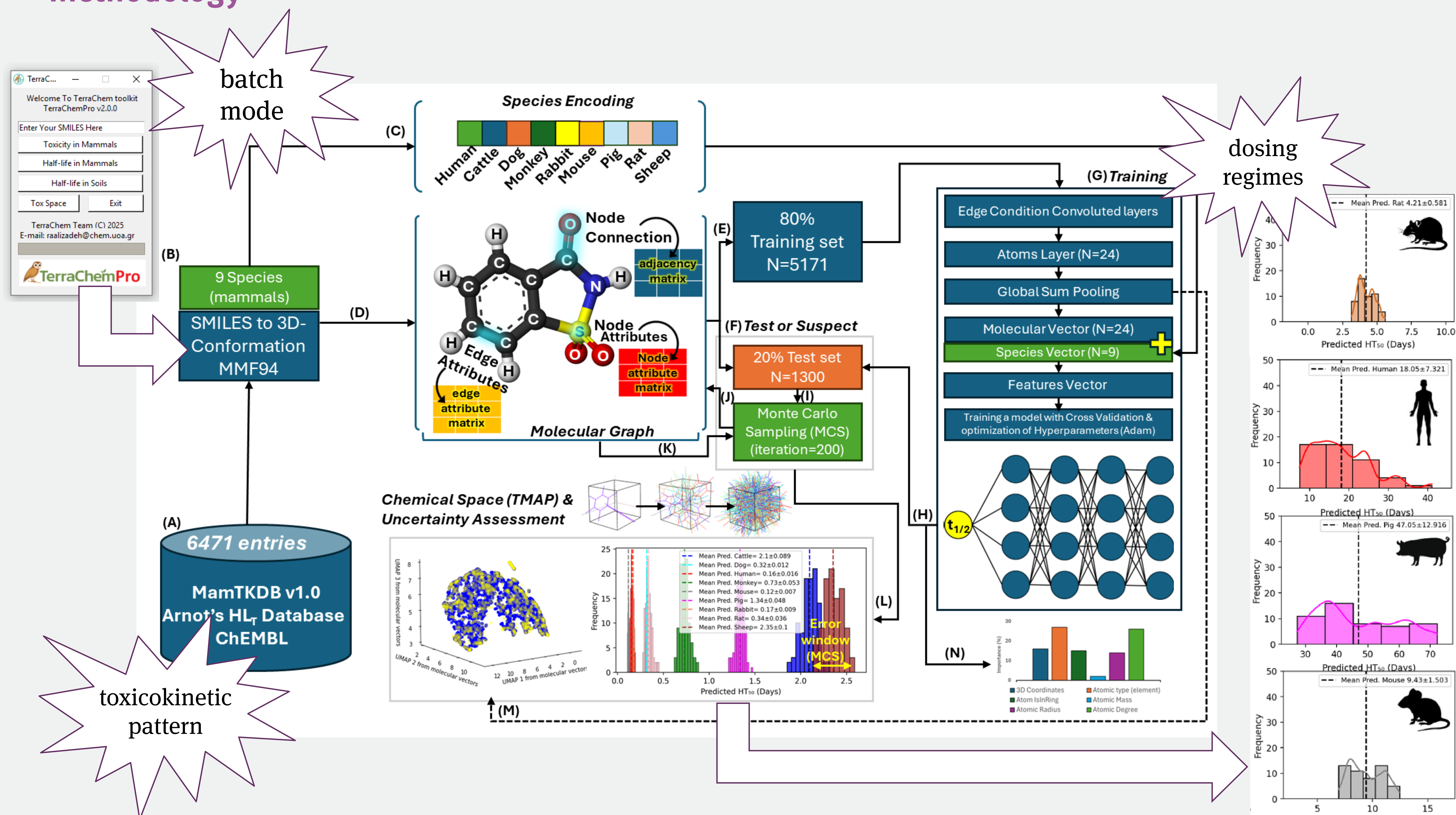


Figure 1: Software features: (A) collection of $elimt_{1/2}$ ^[2-4]; (B) generation of 3D conformers from SMILES; (C) encoding of species data; (D) specification of chemical properties; (E) training and (F) test set; (G) layers of the training procedure; (H) model evaluation; (I) iteration steps; (K) regenerated graph data; (L) estimate standard deviation; (M) UMAP/TMAP-based visualization of learned chemical space; (N) feature importance towards $elimt_{1/2}$.

Outlook -- Open-access software release 2026

- implementation via NORMAN consensus modeling^[5]
- applicable as prioritization tool for regulators & companies to reach sustainability

Literature

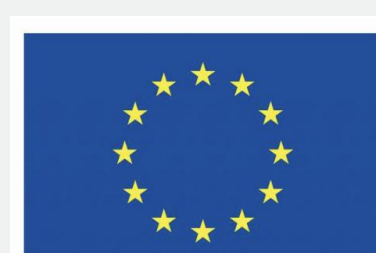
- [1] <https://fclid.ly/airbreathingmammalsl> [2] <https://doi.org/10.1016/j.envint.2021.106592> [3] <https://doi.org/10.1021/acs.jcim.3c02030> [4] <https://pubs.acs.org/doi/10.1021/es4029414> [5] <https://norman-network.com/nds/prioritisation/>

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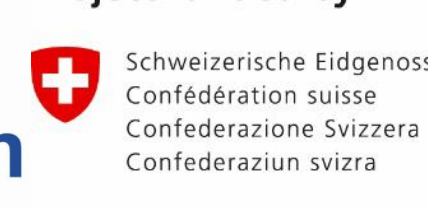
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