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AI and Digital Simulations in Predicting Toxicological Effects of Bioactive Compounds



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Title: Partnership for innovation on the exchange of best practices and the design of joint collaborative initiatives at European level related to the awareness of the effects of contamination on human health

Acronym: INNO-SAFE-LIFE



Introduction



Bioactive compounds from plants, microbes, or synthetic sources are central to drug discovery, nutraceuticals, and cosmetics



Traditional toxicology relies on **in vitro**, **in vivo**, and clinical studies — expensive, time-consuming, ethically complex



AI and computational toxicology offer powerful tools to **predict safety risks** early in the development pipeline



Goal: Use AI and simulations to **reduce cost, time, and animal testing**, while increasing accuracy



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What Are Bioactive Compounds?



Naturally occurring or synthetically produced substances with **biological activity**



Found in **plants, foods, marine organisms, fungi**, etc.



Roles: antioxidants, anti-inflammatory agents, anticancer drugs, antimicrobials



Toxicological risks: genotoxicity, hepatotoxicity, endocrine disruption, etc.

Challenges in Toxicological Assessment



Diverse chemical structures and complex biological pathways



Limited availability of **toxicological data** for novel or rare compounds



Species variability: animal models don't always translate to human outcomes



Regulatory push for **non-animal testing alternatives** (EU REACH, US EPA)

Role of AI in Toxicology



AI algorithms can analyze **large-scale chemical, genomic, and toxicological data**



Machine learning (ML) and deep learning (DL) models identify patterns in structure-toxicity relationships



Predict outcomes such as:

- Mutagenicity
- Carcinogenicity
- Skin sensitization
- Acute/chronic toxicity



Digital Simulations & Modeling Approaches

QSAR (Quantitative Structure–Activity Relationship):

Correlates molecular features with toxicity

PBPK Models (Physiologically Based Pharmacokinetic Modeling): Simulates ADME processes in silico

Molecular docking & dynamics simulations: Assess interaction with biological targets (e.g., DNA, enzymes, receptors)

Virtual organs / Organ-on-a-chip simulations: Model tissue-specific toxicity digitally

AI Tools & Platforms in Use



ProTox-II: Predicts organ toxicity, LD50, pathways



Derek Nexus: Knowledge-based expert system for chemical toxicity



ToxCast & Tox21: Large toxicology databases used to train AI models



DeepTox: Deep learning framework for toxicity prediction from molecular structures



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Applications in Drug and Product Development



Early **screening of natural products** and novel molecules for toxicity red flags



Cosmetic ingredient safety assessment (e.g., skin sensitization prediction)

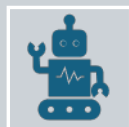


Food and nutraceuticals: Predicting allergenicity and metabolic effects



Environmental impact: Assessing bioaccumulation and ecotoxicity potential

Case Studies



Curcumin analogs: AI-based QSAR models predicted low hepatotoxicity with high anti-inflammatory activity



Marine-derived compounds: Deep learning used to identify neurotoxic potential before extraction and testing



Green tea catechins: In silico metabolism simulation helped assess long-term toxicity risk for supplements

Advantages of AI in Toxicology

Faster predictions than lab testing

Lower costs and fewer animal studies

Ability to handle **large, diverse chemical datasets**

Supports **personalized safety profiling** and precision medicine

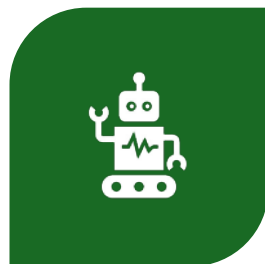




Limitations & Challenges



DATA QUALITY AND AVAILABILITY
LIMIT MODEL ACCURACY



BLACK-BOX NATURE OF SOME AI
MODELS REDUCES
INTERPRETABILITY



NOT ALL BIOLOGICAL RESPONSES
CAN BE SIMULATED (E.G., IMMUNE
INTERACTIONS)



REGULATORY BODIES STILL
CAUTIOUS ABOUT REPLACING
TRADITIONAL METHODS ENTIRELY

Future Trends



Integration of **multi-omics data** (genomics, proteomics, metabolomics) into toxicity prediction



Use of **generative AI** to design safer bioactive compounds



Increasing adoption of **hybrid models** combining AI, lab data, and human clinical insights



Movement toward **regulatory acceptance** of in silico toxicology



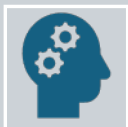
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Conclusion



AI and digital simulations are transforming how we assess the **safety of bioactive compounds**



They offer rapid, cost-effective, and scalable solutions for modern toxicology



With continued advances in **data science, molecular biology, and computing**, these tools will become central to safer, more sustainable innovation in health and wellness