

Artificial Intelligence /Machine Learning for Secondary Metabolite Prediction

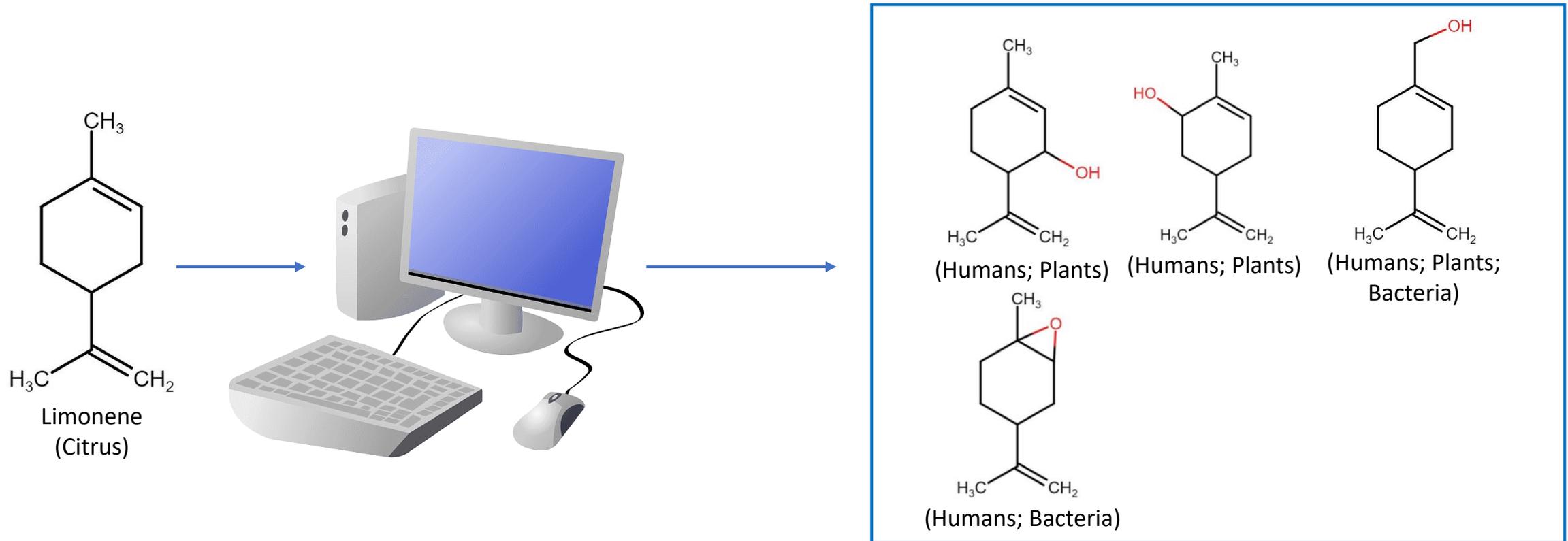
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Corteva Agriscience, Indianapolis, IN, US

Online workshop: Computational Applications in Secondary Metabolite Discovery
March 9, 2021

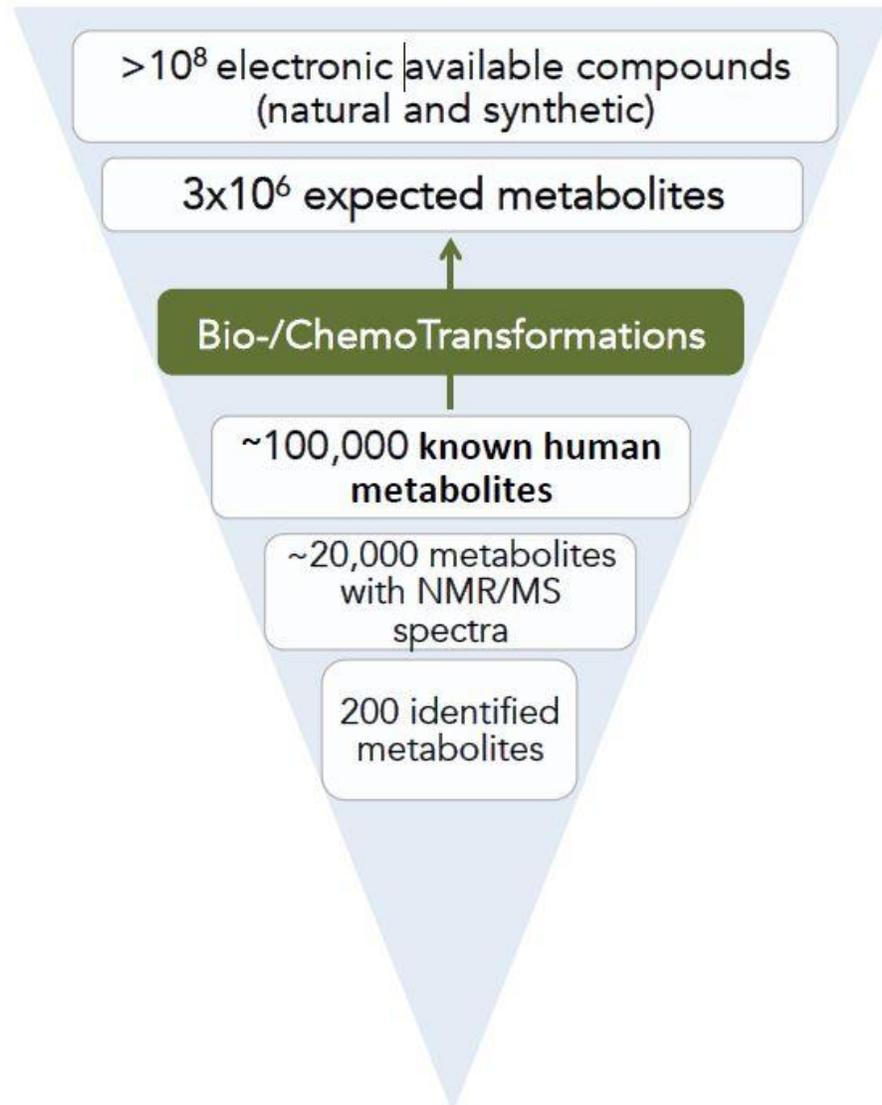
In Silico Metabolism Prediction

Task: Given a small molecule, use computational tools to predict the outcome of its interactions with metabolic enzymes.

Here, we will focus on the structural elucidation of potential metabolites.

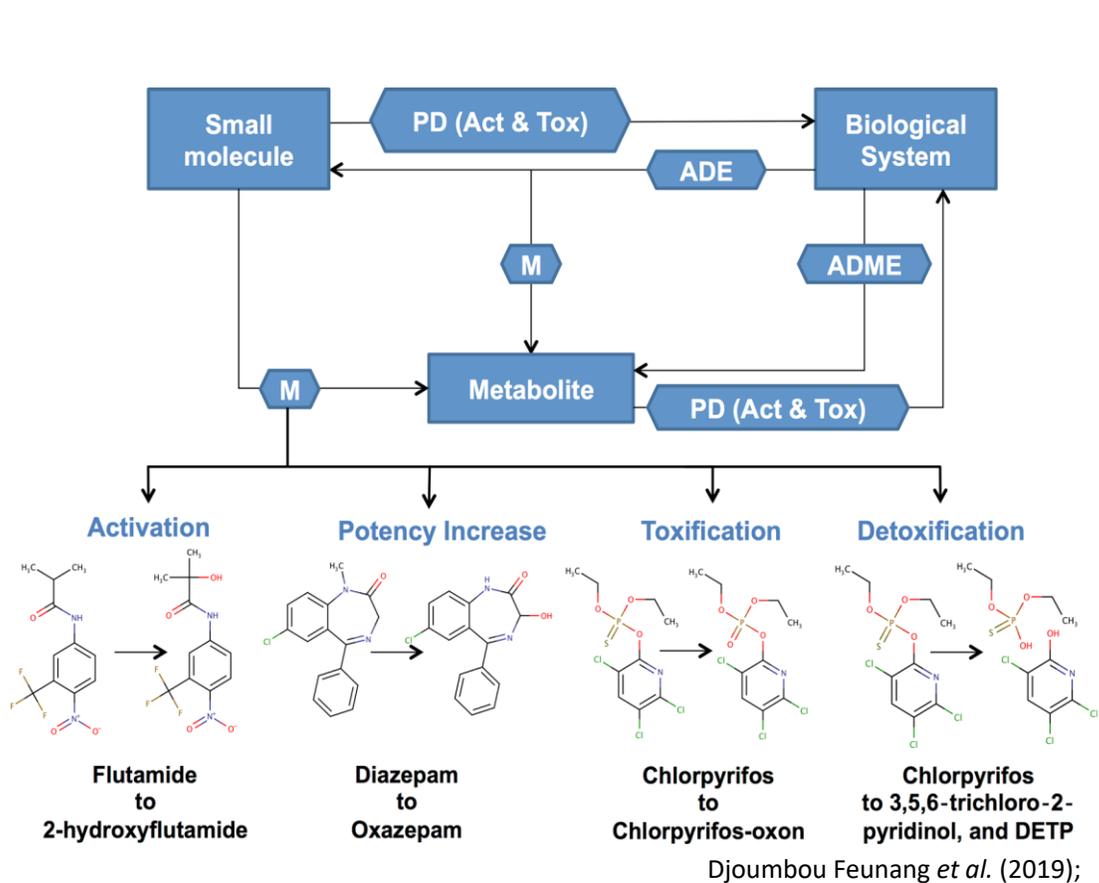


Why is Metabolism so Important?



- Bio/chemo-transformations influence changes of our chemical exposome
- <2% of detectable peaks identifiable in large-scale untargeted Metabolomics
- What are these unknowns?
- How to determine their structures?
- How to determine their activities?
- **Cheminformatics and AI can be used to:**
 - Detect common patterns
 - Predict enzyme/ligand interactions
 - Understand metabolism
 - Generate biologically feasible structures through simulated reactions

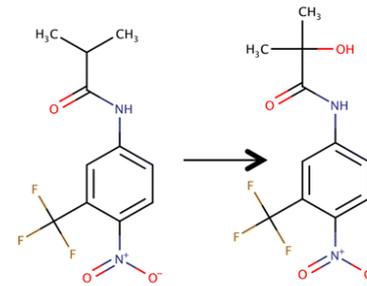
Why is Metabolism so Important?



Influence of metabolism on a xenobiotic's pharmacodynamics (PD), including pharmacological activity (Act), and toxicological effects (Tox).

Pro/soft-drug design

Activation

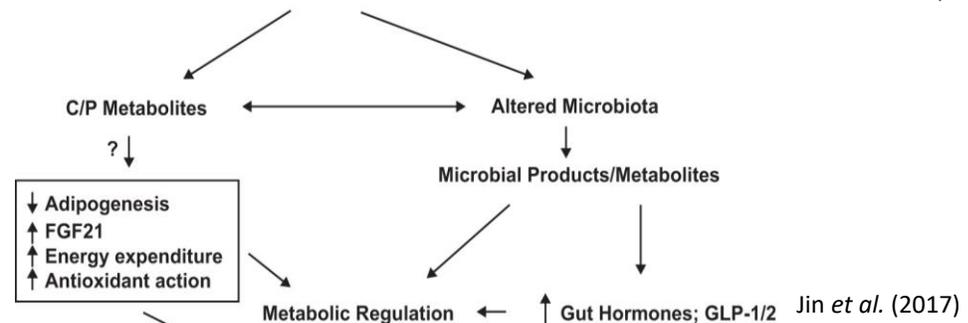


Exposure science

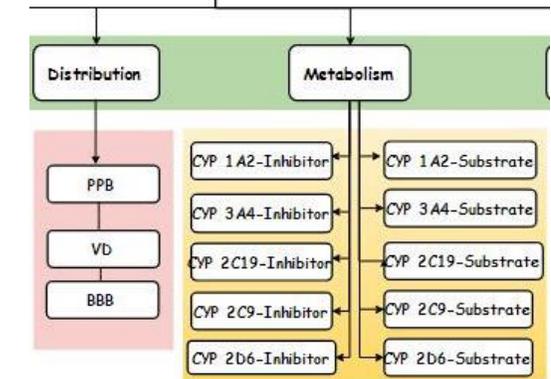


Nutrition science

Curcumin/Polyphenols (C/P)

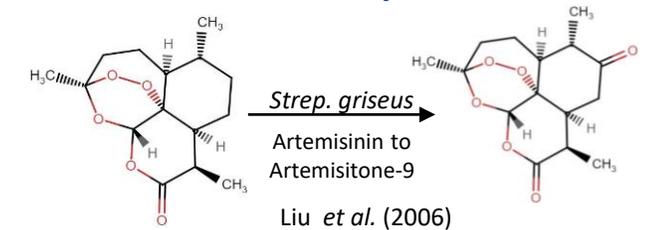


ADMETlab ADMET profiling

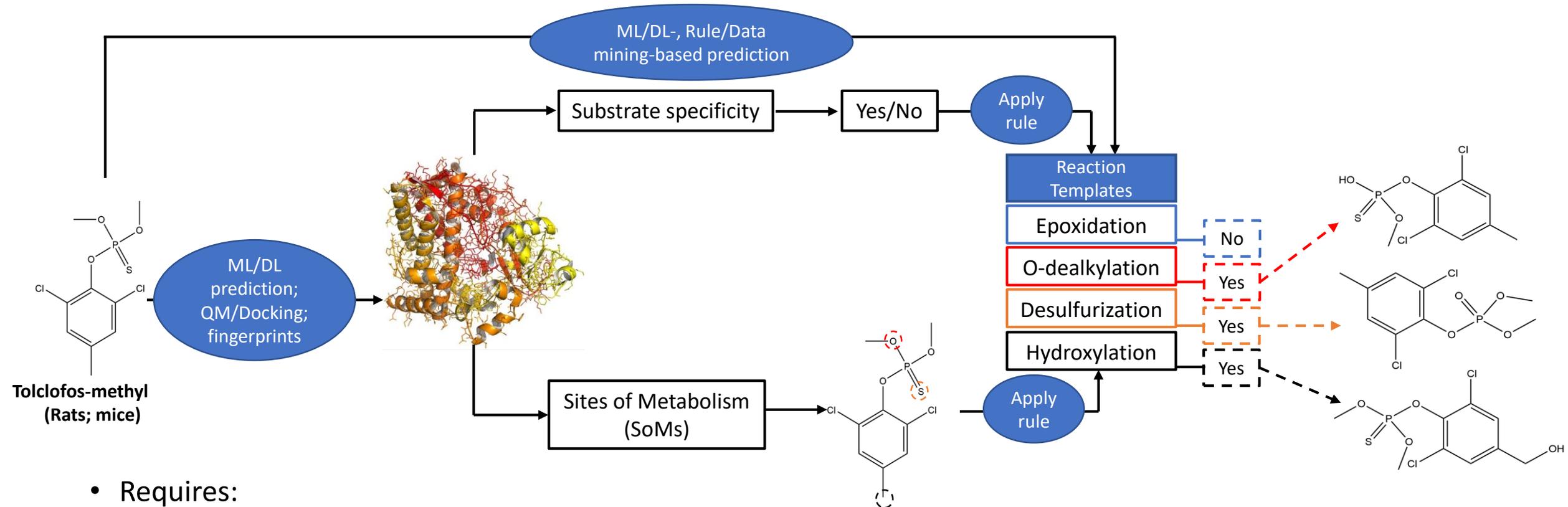


<http://admet.scbdd.com/>

Microbial biosynthesis



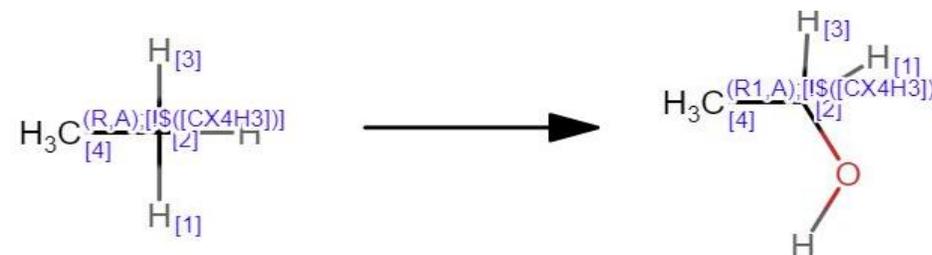
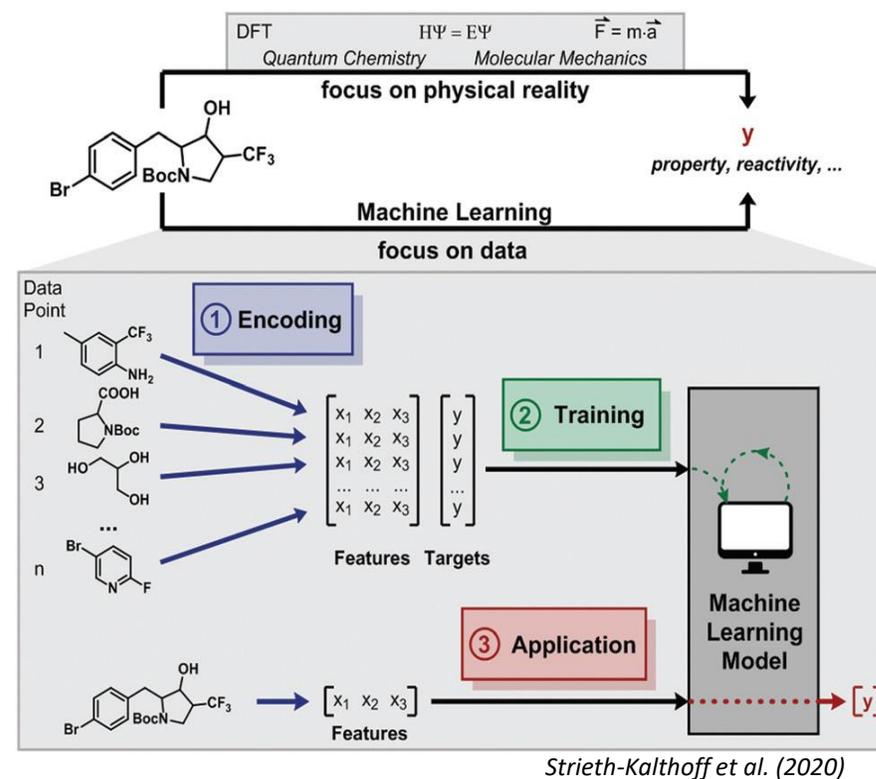
Approaches for *In silico* Metabolism Prediction



- Requires:
 - Module to predict (and rank/score) SoMs, enzyme-substrate selectivity, or reaction groups
 - Library of reaction templates to apply or select (via prediction) from
 - Modules are usually specific (chemical/enzyme classes), or comprehensive (whole species)
- Prediction approaches can be ligand- or structure-based
- Some tools include: ML/DL-based (MetaTrans, Glory), Rule-based (MetabolExpert), Hybrid (BioTransformer, Meteor Nexus)

Ligand-based Prediction

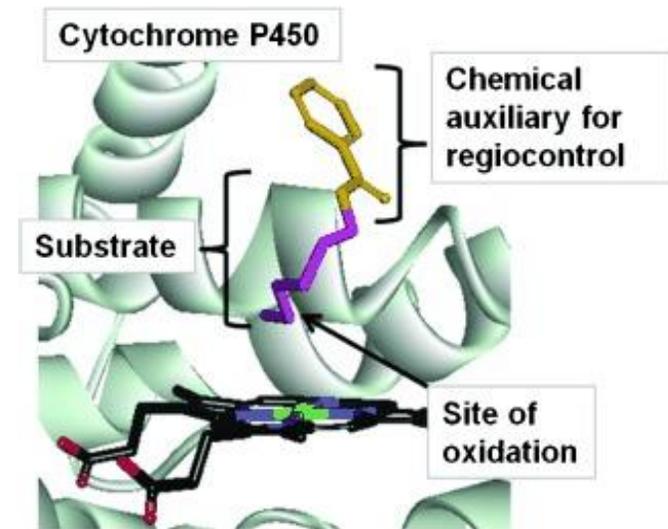
- Use structures of ligands/non-ligands to predict specificity, accessibility to enzymes
- Some methods include, among others:
 - QSAR/QSMR (SoM, BoM, ESS)
 - Data mining/fingerprints (SoM)
 - Substructures/Rules (ESS, Reaction)
- Advantages:
 - Speed
 - Seem to perform as well as structure-based methods
- Disadvantages:
 - Approaches/models/rule bases tend to not work on novel chemistries
 - Data quantity and quality is a limiting factor



Hydroxylation of methyl carbon adjacent to aliphatic ring

Structure-based Prediction

- Explicitly model the interaction within the enzyme's binding pocket
- Help predicting sites of metabolism (SoMs)
- Some methods include, among others:
 - Protein-ligand docking
 - Molecular dynamics
- Advantages
 - Detailed study of the enzyme-substrate interaction
- Disadvantages
 - High computational power to model structural flexibility



Ménard *et al.* (2012)

Examples of *In Silico* Metabolism Prediction Tools

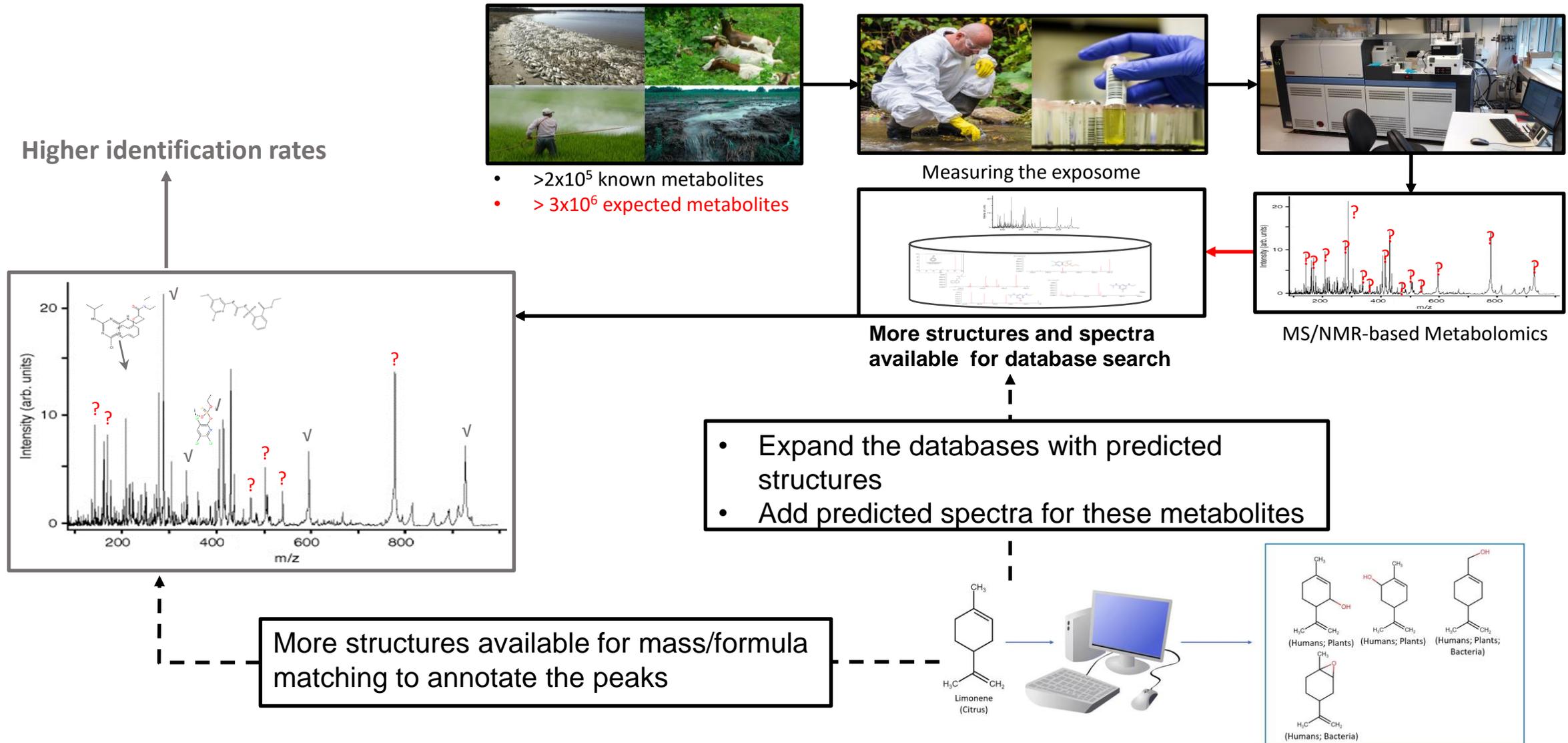
Based on their coverage, *in silico* metabolism prediction tools can be classified as:

1. **Specific tools**, which apply to simple biological systems (e.g. a small set of enzymes), and often, cover a rather small chemical space.
2. **Comprehensive tools**, which apply to a more diverse and larger set of enzymes, species, and cover a larger chemical space.

Prediction Tool	Accessibility	Spectrum	Approach	Methods	Availability
BioTransformer	Web server / Command line	Comprehensive	Ligand-based	Machine Learning Expert system	Open source
Meteor Nexus	GUI-based standalone	Comprehensive	Ligand-based	Machine Learning Expert system	Commercial
GLORYx	Web server	Comprehensive	Ligand-based	Machine Learning	Freely available
SmartCYP	Web server	Specific	Combined	Machine Learning	Freely available
MetaSite	GUI-based standalone	Specific	Combined	Machine Learning	Commercial
MetabolExpert	GUI-based standalone	Specific	Ligand-based	Expert system	Commercial

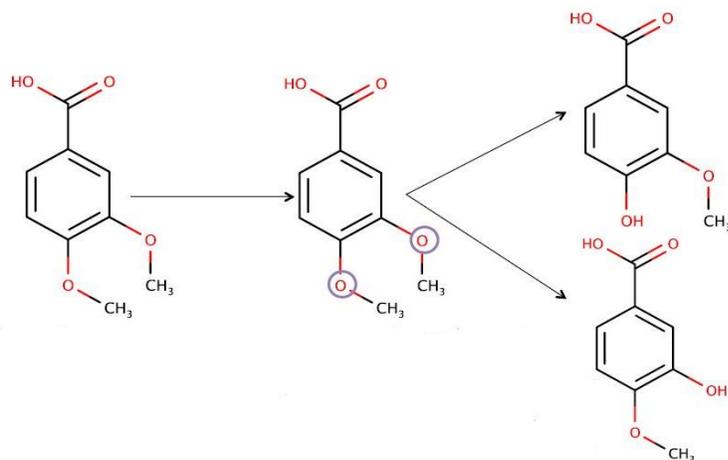
Examples of tools for metabolite prediction

Enhancing Metabolite Discovery and Identification

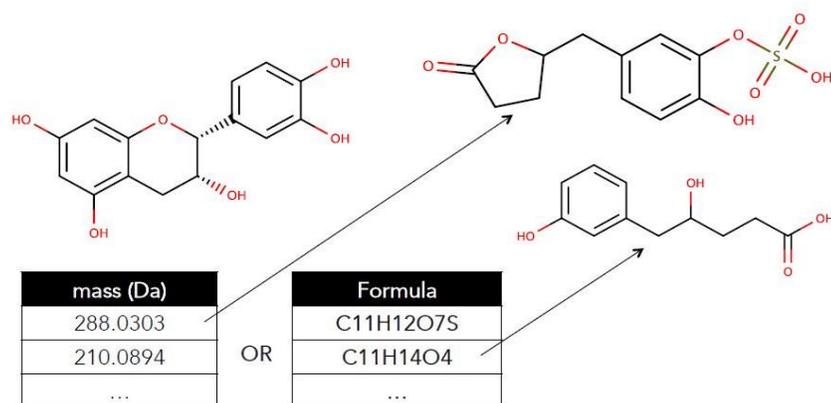


BioTransformer: Open Source for Metabolite Identification

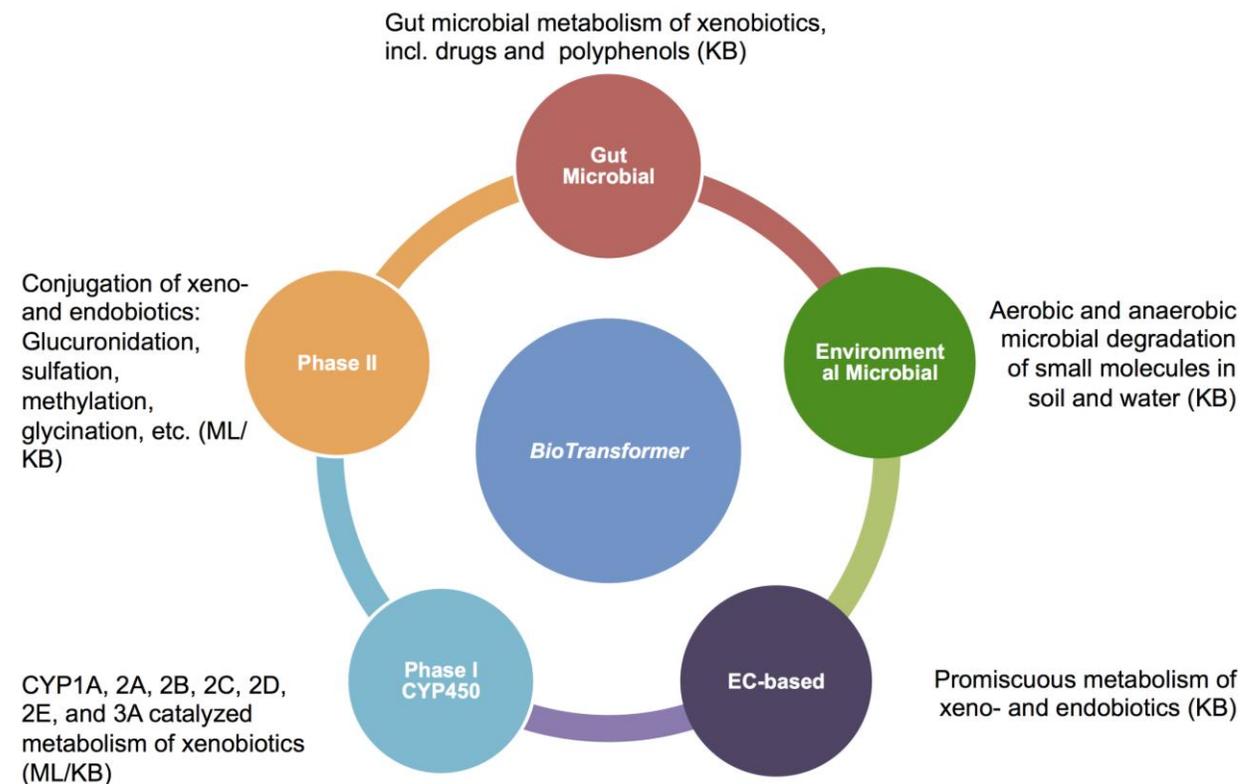
BioTransformer Metabolism Prediction Tool



BioTransformer Metabolite Identification Tool

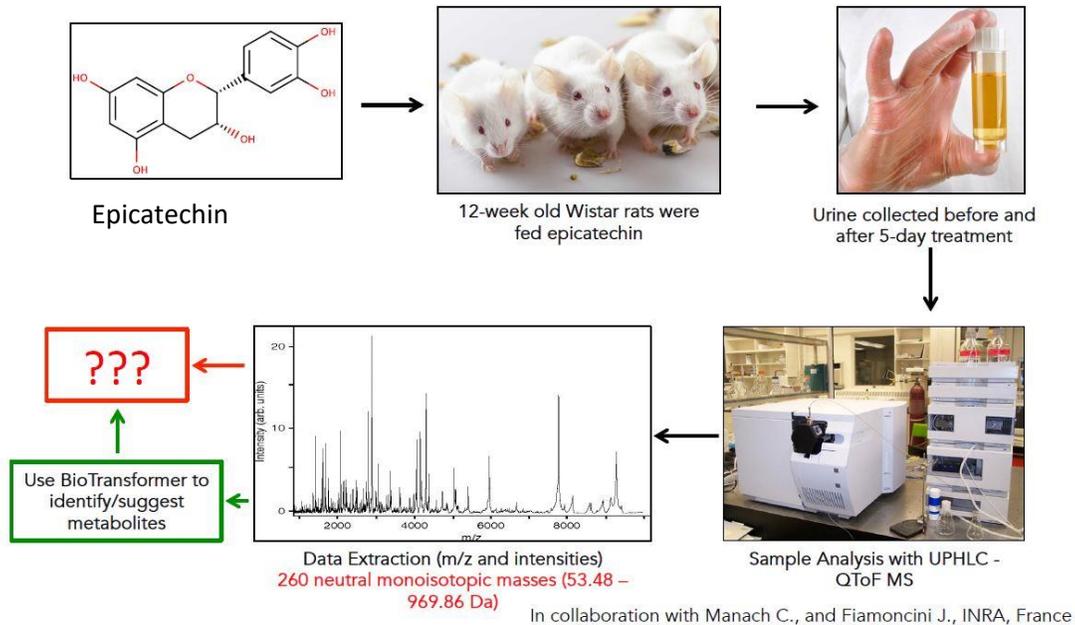


Djoumbou Feunang *et al.* (2019); J. Cheminf.; DOI [10.1186/s13321-018-0324-5](https://doi.org/10.1186/s13321-018-0324-5)

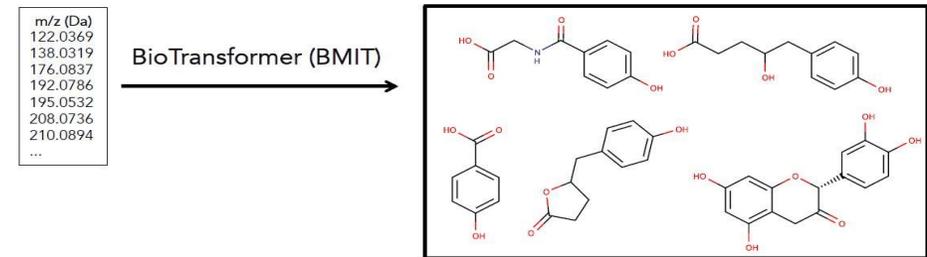


- **Djoumbou-Feunang *et al.* J Cheminform. 2019 Jan 5;11(1):2**
- Open Source: bitbucket.org/djoumbou/biotransformer
- Docker: <https://hub.docker.com/r/djoy2018/biotransformer-cl>
- RESTful web server: www.biotransformer.ca

BioTransformer: Open source for Metabolite Identification

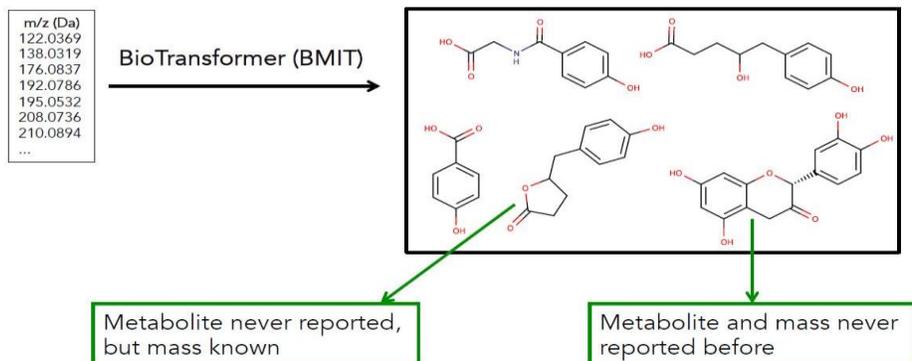


- Literature mining revealed 56 single- and multi-step epicatechin metabolites corresponding to 37 unique masses
 - 11 out of 37 masses were measured in our experimental study

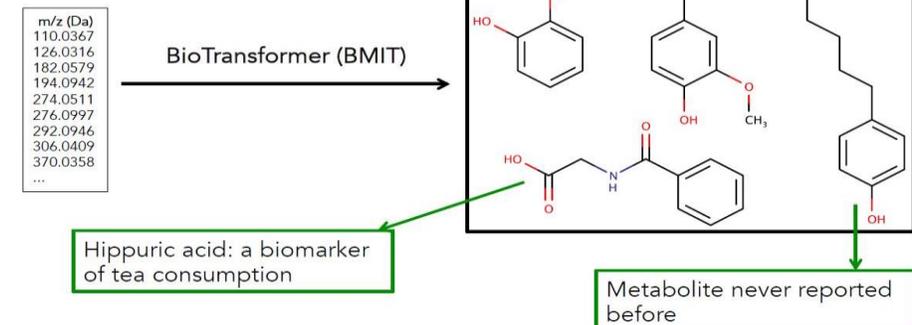


- BioTransformer (BMIT) suggested 37 compounds (20 masses)
 - 22/37 matched the 11 unique and previously reported masses
 - 18/22 compounds were confirmed in previous reports

- BioTransformer (BMIT) was used to identify metabolites matching the 26 masses not measured in our study



- BioTransformer (BMIT) suggested 19 new compounds
 - 11 compounds match previously reported masses
 - 15 compounds match masses observed exclusively in our study
 - Experimental validation needed



- BioTransformer (BMIT) suggested 28 compounds (19 masses)
 - 21 previously reported metabolites matching 18 unique masses
 - Experimental validation needed for the 7 new compounds

Challenges in Metabolism Prediction

- Challenges in understanding metabolism lead to lower prediction accuracy
 - ML tools can rank SoMs with high accuracy, but not the resulting biotransformations
 - Knowledge-driven approaches achieve higher recall, but suffer from combinatorial explosion
- Limited accessibility/availability of tools
 - Most tools developed in academia are available only via web-server, raising confidentiality issues for most potential users
 - Restricted shareability of data generated by commercial tools
- The limited availability of high-quality data
 - Crucial for achieving better accuracy, higher coverage, and larger applicability domains
 - Needed to derive more reaction rules
 - However, data curation (Sites of metabolism, reaction annotation) is tedious and expensive

Outlook

- Data sharing
 - Increase the amount of high-quality, publicly available, curated, and downloadable data (substrate-product, Sites of Metabolism). E.g.: MetXBioDB, PubChem, XMetDB, etc.
 - Develop publicly accessible databases of predicted metabolites (e.g.: BioTransformerDB)
 - Community-wide efforts needed
- Novel, and innovative AI approaches:
 - Seq2Seq transformer architectures have proven applicable for end-to-end learning-based method, with results comparable to existing tools (MetaTrans: Litsa *et al.* 2020)
 - Could improve accuracy, while bypassing manual rule design
 - Bonds of Metabolism (BoMs) seem to provide a better description of reaction centers, leading to higher accuracy, compared to SoMs (Upcoming - Tian, et al. (2021))
- Open source communities
 - Provide means for easier user feedback loops; valuable for improving prediction tools
 - Provide developers opportunities to improve software tools, in an agile, continuous manner
 - Successful projects include (BioTransformer, Chemistry Development Kit, Knime)

Thank You

- The organizing committee:
 - Fidele Ntie-Kang
 - J. Ludwig Muller
- The Wishart Lab @UofAlberta, Canada
- Corteva Agriscience
- The BioTransformer community
- The listeners

- To learn more about BioTransformer:
 - *HS03: In Silico Prediction and Identification of Metabolites with BioTransformer : Enabling Secondary Metabolite Discovery; March 10, 2021*

