



Computational Tools for the Identification of Unknowns

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**3rd International Electronic Conference on
Metabolomics**

Nov. 15-30, 2018

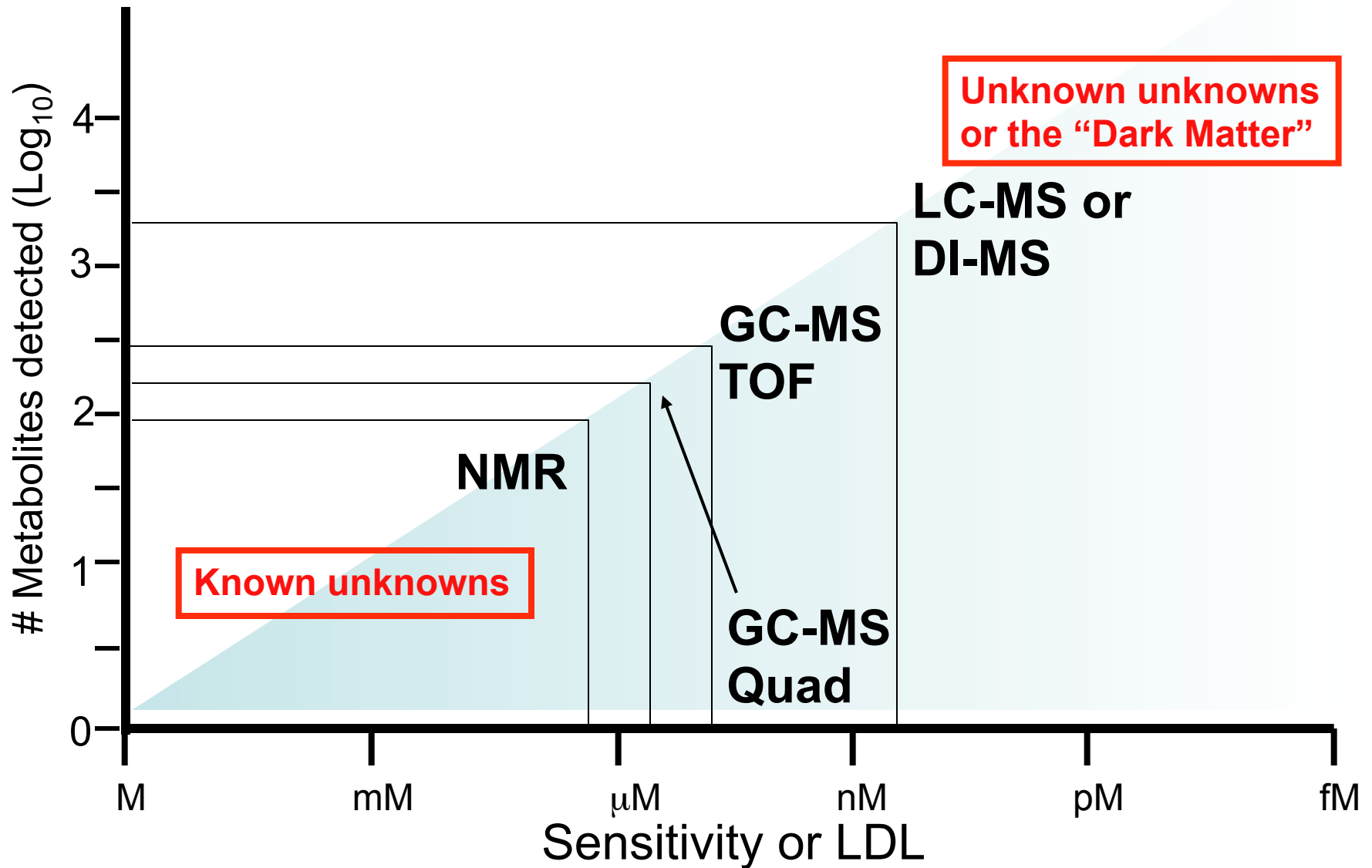
What We Don't Know

- ***“...there are known unknowns; that is to say we know there are some things we do not know. But there are also unknown unknowns – the ones we don't know we don't know.”***

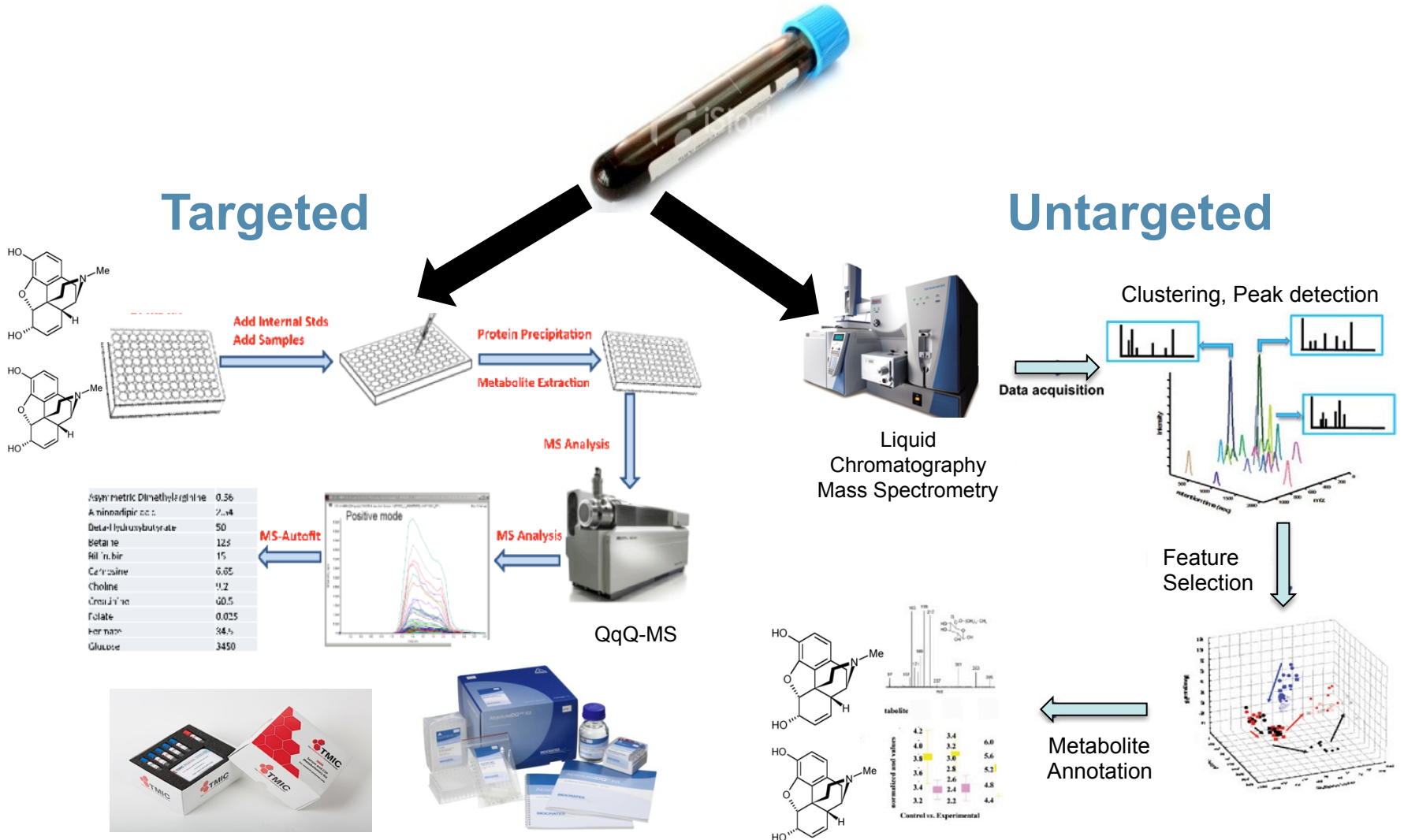


Donald Rumsfeld, US Secretary of Defense - Nov. 2001

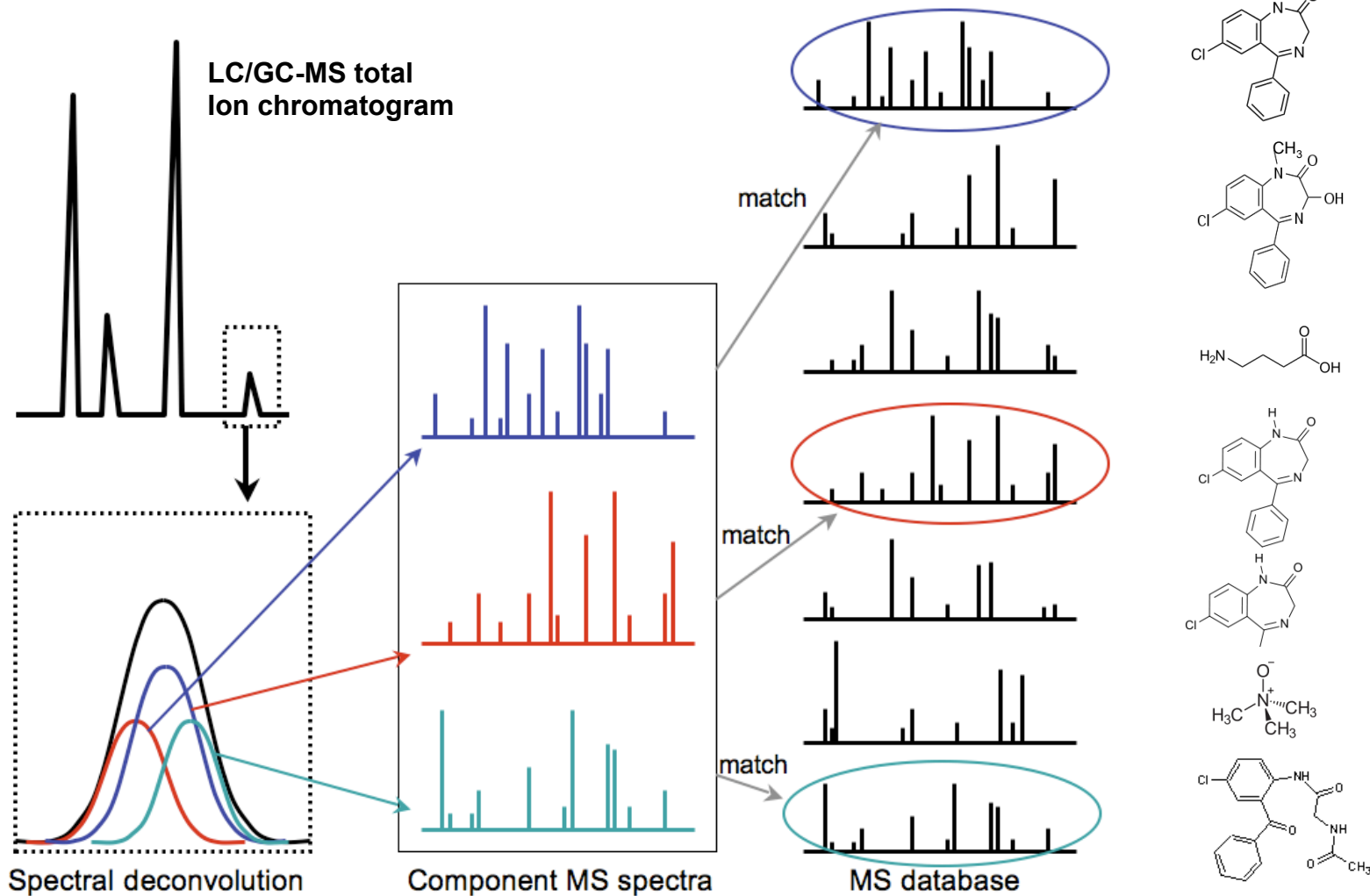
Technology & Sensitivity



2 Routes to Metabolomics



Untargeted MS Compound Identification



Levels of Metabolite ID for Untargeted Metabolomics

- 4 levels of metabolite identification
- **Level 1 - Positively identified compounds**
 - Confirmed by MS/MS match and RT match to an actual/authentic standard
- **Level 2 - Putatively identified compounds**
 - Match to EI-MS + RT or MS/MS + RT from a reference database
- **Level 3 - Compounds putatively identified via molecular formula or m/z matching**
 - Match to high resolution m/z and nothing else
- **Level 4 - Unknown compounds**

Compound Identification (Formula Matching – Level 3 ID)

The screenshot shows the ChemSpider website. At the top, there is a navigation bar with links for Home, About us, Web APIs, Help, and Sign in. The ChemSpider logo is prominently displayed, along with the tagline "Search and share chemistry". A search bar is located in the top right corner. Below the navigation bar, there are tabs for "Simple", "Structure", "Advanced", and "History". The main content area features a "Search ChemSpider" section with a search input field and a search button. Below the search field, there is a list of search criteria: "Systematic Name, Synonym, Trade Name, Registry Number, SMILES, InChI or CSID". To the right of the search section, there is an advertisement for "Empower 3" chromatography software. On the left side, there is a "What is ChemSpider?" section and a "Search by chemical names" section with a list of search criteria: "Systematic names", "Synonyms", "Trade names", and "Database identifiers".

68 million chemicals

The screenshot shows the PubChem website. At the top, there is a navigation bar with links for Databases, Upload, Services, Help, more, and Today's Statistics. The PubChem logo is prominently displayed. Below the logo, there are three tabs: "BioAssay", "Compound", and "Substance". A search input field is located below the tabs, with a "Go" button and "Limits Advanced" link. Below the search field, there is a "Try the PubChem Search Beta" section. At the bottom, there is a notice about the retirement of PubChem BioAssay Tools. On the right side, there is a sidebar with various tools and services: "BioAssay Tools", "Structure Search", "3D Conformer Tools", "Structure Clustering", "Classification", "Upload", "Download", and "PubChem FTP".

96 million chemicals

Compound Identification (Spectral Matching – Level 2 ID)

14,009 “real” compounds
72,036 “real” spectra
~150,000 total compounds

213,019 MS spectra
75,270 compounds

Welcome to MoNA!

MassBank of North America (MoNA) is a metadata-centric, auto-curating repository designed for efficient storage and querying of mass spectral records. It intends to serve as the framework for a centralized, collaborative database of metabolite mass spectra, metadata and associated compounds. MoNA currently contains over 200,000 mass spectral records from experimental and in-silico libraries as well as from user contributions.

MoNA has recently been redesigned, with significant improvements to server-side architecture, query structure, and search speed. We are actively improving and adding features, so please be patient as functionality is added. If you notice any major issues, feel free to report them using the issue tracker link below.

Name	Avg. Score	Spectra
Tobias Schulze	★★★★★	2,867
Stephan Beisen	★★★★★	58
Nikolaos Thomaidis	★★★★★	1,492
Martin Krauss	★★★★★	622
Emma Schymanski	★★★★★	11,656
Megan Showalter	★★★★★	31

Other Resources for MS/MS Spectral Matching

- **HMDB** – 302,219 spectra, 114,100 cmpds
- **mzCloud** – 191,722 spectra, 8304 cmpds
- **NIST17 MS/MS** – 652,475 spectra, 14,351 cmpds
- **MassBank** – 28,185 spectra, 11,500 cmpds
- **Wiley LC-MSⁿ** – >10,000 spectra, 4500 poisons
- **ReSpect** – 9017 spectra, 3595 cmpds
- **GNPS** – 221,000 spectra, 18,163 cmpds

How Well Do We Do?

- **Untargeted LC-MS of human biofluids** – 100-250 compounds positively ID'd (level 2) out of 10,000+ features (1%), 700-1000 tentatively ID'd (level 3) out of 10,000+ features (8%)
- **Untargeted LC-MS of river water** – 649 compounds identified (level 1 or 2) out of 8535 features (8%) (*Schymanski et al. Anal Bioanal Chem (2015) 407:6237–6255*)

Overall we are doing pretty badly

Why Are We Doing So Badly?

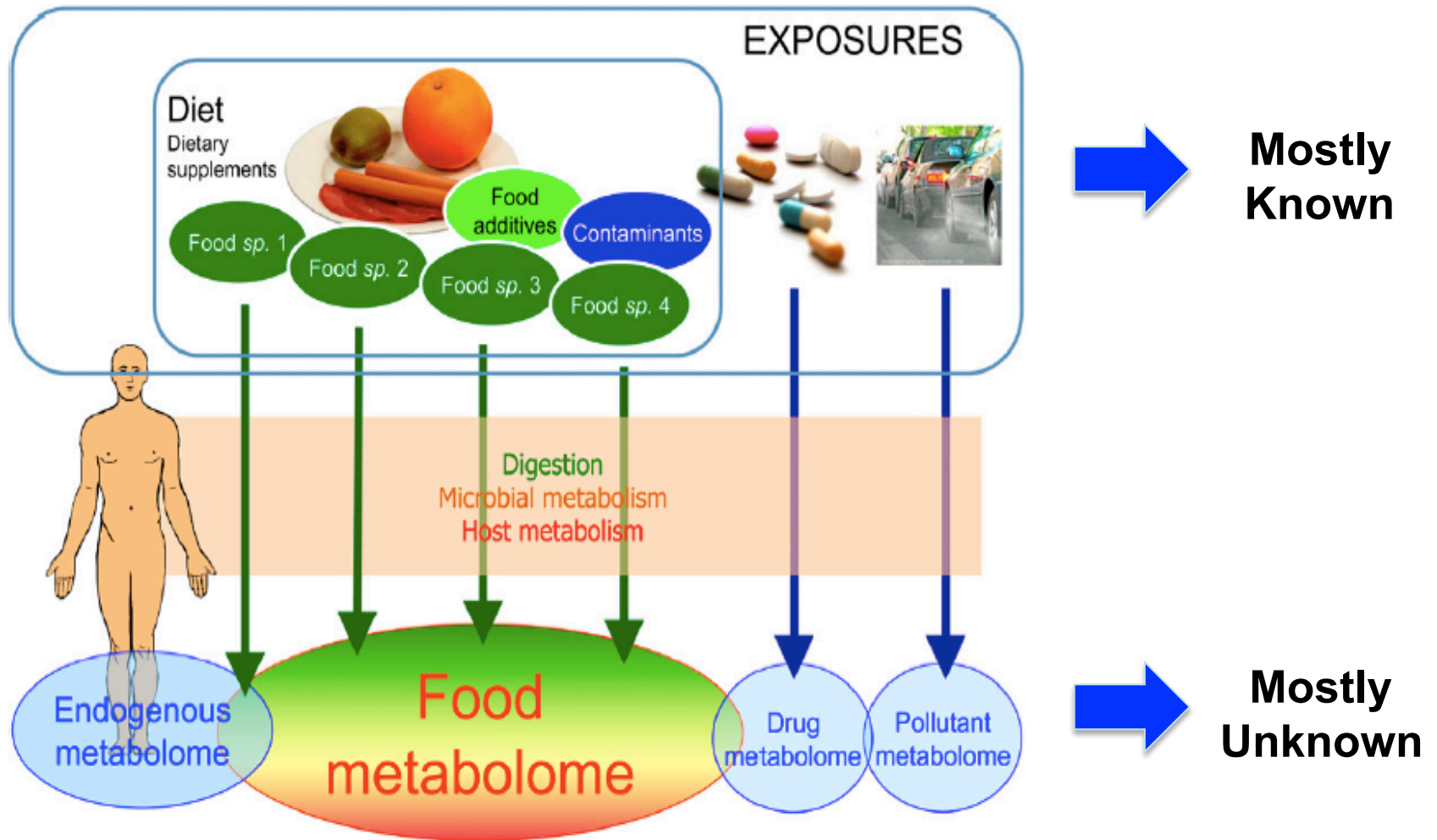
- **32,000,000,000** chemical formulae (<2000 Da)
- **2,400,000,000** chemically feasible formulae
- **96,500,000** chemicals in PubChem
- **1,500,000** LC/GC-MS spectra collected on ~15 different platforms
- **80,000** chemicals with EI-MS spectra
- **~20,000** chemicals with high resolution MS spectra
- **~1500** chemicals that are biologically relevant



Why Are We Doing So Badly?

- Using larger databases (PubChem, ChemSpider) and m/z matching is leading to many, many false positives
- <0.2% of compounds in PubChem or ChemSpider have ever left the laboratory or are likely to be found in humans or in the environment (**Bigger isn't better**)
- Only 1500 “meaningful” chemicals are routinely available from vendors (**Synthetic chemistry is hard**)
- Enormous resources going to collect lots of MS spectra on a tiny number of chemicals (**Measuring the same thing over doesn't make the problem go away**)
- Most of the unknowns are not in PubChem or Chemspider, or anywhere else (**Where are they?**)

What Are These Unknowns?



What To Do?

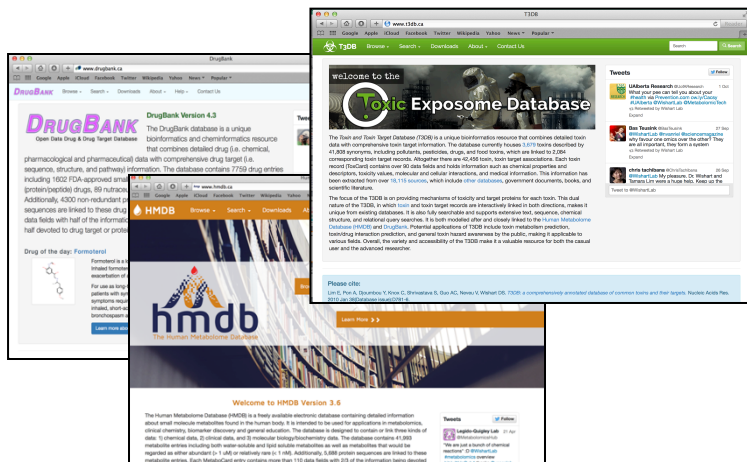
- Obtain or synthesize all commonly available xenobiotics (HPVs, drugs, pollutants, foods, etc.), prepare or synthesize their metabolites and collect their NMR, LC-MS and GC-MS spectra
COST: 5 million cmpds X \$1000/cmpd = \$5 billion
- **OR**
- Do this entire exercise computationally
COST - 5,000,000 cmpds X \$0.10/cmpd = \$500,000

***In Silico* Metabolomics**

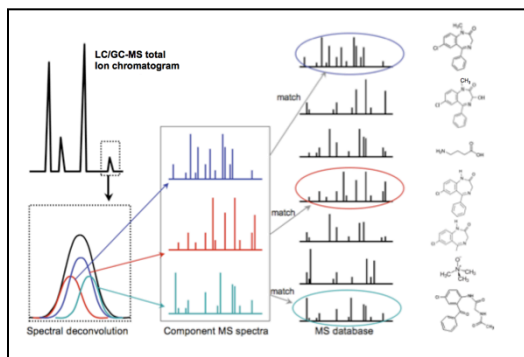
- **An emerging concept to facilitate metabolite ID of unknown unknowns**
- **Realization that all metabolites will never be synthesized or isolated and most will never have reference MS/MS or NMR spectra collected**
- **Based on *in silico* prediction of biologically feasible metabolites**
- **Based on *in silico* prediction of observables (RI, RT, NMR spectra, IR, CCS, MW, MS/MS spectra)**

In Silico Metabolomics

HMDB/DrugBank/T3DB

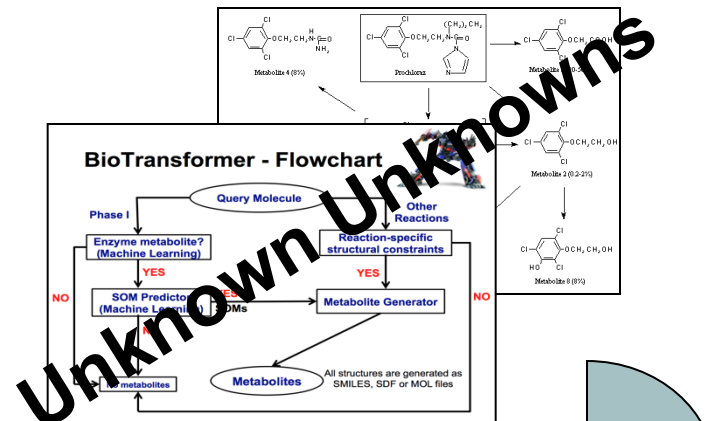


Known compounds (250,000)



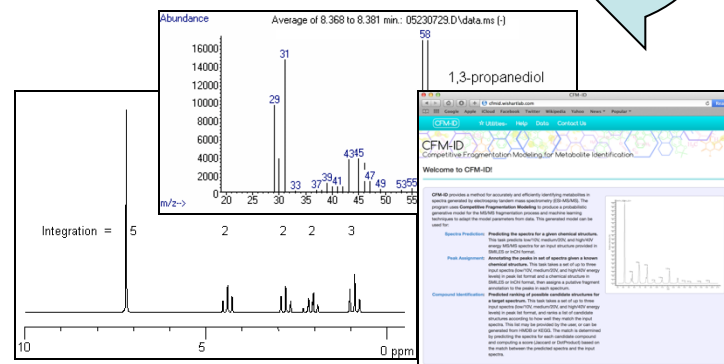
Match observed spectra to predicted spectra to ID compounds

BioTransformer



Predicted biotransformations (250,000 --> 5,000,000)

CFM-ID/NMRPred



Predicted MS/MS, NMR, GC-MS spectra of knowns + biotransformed

What Is Known?

UofA Metabolomics Databases



www.hmdb.ca



www.drugbank.ca



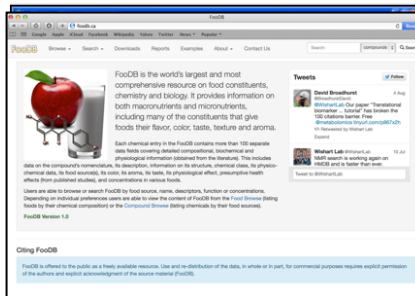
www.ymdb.ca



www.phenol-explorer.eu



www.ecmdb.ca



www.foodb.ca



www.cowmetdb.ca



www.t3db.ca



www.smpdb.ca



www.csfmetabolome.ca



www.serummetabolome.ca



www.urinemetabolome.ca

The **New** Human Metabolome Database (HMDB)

The image shows two overlapping screenshots of the Human Metabolome Database (HMDB) website. The top screenshot shows the main homepage with the HMDB logo and navigation options. The bottom screenshot shows a detailed view of metabolites, listing their names, structures, formulas, average masses, mono masses, and biofluid locations.


HMDB ID	Name	Structure	Formula	Average Mass	Mono Mass	Biofluid Location
HMDB000001 332-80-9	1-Methylhistidine	<chem>CN1C=NC(=C1)C(=O)N</chem>	C ₇ H ₁₁ N ₃ O ₂	169.1811	169.085126611	Blood Cerebrospinal Fluid (CSF) Saliva Urine
HMDB000002 109-76-2	1,3-Diaminopropane	<chem>NCCN</chem>	C ₃ H ₁₀ N ₂	74.1249	74.08439833	Blood Urine
HMDB000006 600-18-0	2-Ketobutyric acid	<chem>CC(=O)C(=O)O</chem>	C ₄ H ₆ O ₃	102.0886	102.031694058	Blood Cerebrospinal Fluid (CSF) Saliva Urine
HMDB000008 600-15-7	2-Hydroxybutyric acid	<chem>CC(O)C(=O)O</chem>	C ₄ H ₈ O ₃	104.1045	104.047344122	Blood Cerebrospinal Fluid (CSF) Saliva Urine

<http://www.hmdb.ca>

- HMDB 4.0 has 114,100 “quantified”, “detected”, “expected” and “predicted” metabolites (3X more than version 3.0)
- HMDB 3.0 had 442 biological pathways, HMDB 4.1 has 48,627 (100X more than version 3.0)
- New version has >500,000 MS/MS & GC-MS spectra, 3900 NMR spectra
- New version has 6800 metabolite-SNP associations, 2500 metabolite-drug associations and 2900 metabolite-age/gender associations
- 78,000 new lipids/peptides to be added in late 2018 – total = 192,000

The **New** Drug Database (DrugBank v. 5.0)

The screenshot shows the DrugBank website interface. The top navigation bar includes 'DRUGBANK', 'Browse', 'Search', 'Downloads', 'About', 'Help', 'Blog', and 'Contact Us'. A search bar contains the text 'Tylenol'. Below the search bar are buttons for 'Drugs', 'Targets', 'Pathways', and 'Indications'. The main content area displays the DrugBank logo and a brief description: 'The DrugBank database is a unique bioinformatics and cheminformatics resource with comprehensive data on drugs, drug actions and drug targets. The latest release contains 2,533 approved pharmaceuticals, 5,700 experimental drugs, 3,850 drug metabolites, 1,360 drug targets, 6,000 MS+NMR spectra, and 5,130 unique drug targets. Each drug entry contains 215 data fields.' Below this, a detailed view of Acetaminophen is shown, including its name, accession number (DB00316), type (Small Molecule), groups (Approved), description, chemical structure, and synonyms.

Name	Acetaminophen
Accession Number	DB00316 (APR000252)
Type	Small Molecule
Groups	Approved
Description	Acetaminophen, also known as paracetamol, is commonly used for its analgesic and antipyretic effects. Its therapeutic effects are similar to salicylates, but it lacks anti-inflammatory, antiplatelet, and gastric ulcerative effects.
Structure	 3D Download Similar Structures
Synonyms	4-(Acetylamino)phenol 4-acetamidophenol 4'-hydroxyacetanilide

- A comprehensive database of drugs, drug actions and drug targets
- 2533 small molecule drugs
- >5700 experimental drugs
- Detailed ADMET, MOA and pharmacokinetic data
- >3850 drugs with metabolizing enzyme data
- >1360 drug metabolites
- >6000 MS+NMR spectra
- >5130 unique drug targets
- 215 data fields/drug
- **Published on Jan. 1, 2018**

<http://www.drugbank.ca>

The Food Database (FoodDB)

The screenshot shows the FoodDB website interface. The top navigation bar includes 'Browse', 'Search', 'Downloads', 'Reports', 'Examples', 'About', and 'Contact Us'. The main content area features a large image of a red apple and a glass of milk, with a chemical structure overlaid. Text describes FoodDB as the world's largest and most comprehensive resource on food constituents, chemistry, and biology. It provides information on both macronutrients and micronutrients, including many of the constituents that give foods their flavor, color, taste, texture and aroma. Each chemical entry in the FoodDB contains more than 100 separate data fields covering detailed compositional, biochemical and physiological information (obtained from the literature). This includes data on the compound's nomenclature, its description, information on its structure, chemical class, its physico-chemical data, its food source(s), its color, its aroma, its taste, its physiological effect, presumptive health effects (from published studies), and concentrations in various foods. Users are able to browse or search FoodDB by food source, name, descriptors, function or concentrations. Depending on individual preferences users are able to view the content of FoodDB from the Food Browser (listing foods by their chemical composition) or the Compound Browser (listing chemicals by their food sources). FoodDB Version 1.0

Citing FoodDB

FoodDB is offered to the public as a freely available resource. Use and re-distribution of the data, in whole or in part, for commercial purposes requires explicit permission of the authors and explicit acknowledgment of the source material (FoodDB).

Description
Angelica is a genus of about 100 species of tall perennial herbs in the family Apiaceae, native to temperate and subarctic regions of the Northern Hemisphere, reaching as far north as Iceland and Lapland. They grow to 1-3 m tall, with large bipinnate leaves and large compound umbels of white or greenish-white flowers. Some species can be found in purple moor and rush pastures.

Picture

Classification

Group	Herbs and Spices
Sub-Group	Herbs

External Links

ITIS ID	Not Available
Wikipedia ID	Angelica

Composition

Compounds: Show 10 entries

Compound	Structure	Content Range	Average	Reference Type	Reference
<input type="text" value="Search Compound"/>				<input type="text" value="Search Reference"/>	

- A comprehensive food composition database (more than polyphenols)
- 28,771 compounds
- 718,405 concentration values for 722 raw/processed foods
- 31,791 references
- 1435 cmpds with health effects
- 2692 cmpds w flavour attributes
- 2000+ reference MS/NMR spectra
- Structure & text searches
- >100 data fields/compound
- *Publicly released on Jan. 1, 2018, manuscript being prepared*

www.foodb.ca

The Toxic Exposome Database (T3DB)

The screenshot shows the T3DB website. The top part is the homepage with a 'welcome to the Toxic Exposome Database' banner. Below the banner is a paragraph describing the database: 'The Toxin and Toxin Target Database (T3DB) is a unique bioinformatics resource that combines detailed toxin data with comprehensive toxin target information. The database currently houses 3,679 toxins described by 41,808 synonyms, including pollutants, pesticides, drugs, and food toxins, which are linked to 2,084 corresponding toxin target records. Altogether there are 42,456 toxin, toxin target associations. Each toxin record (ToxCard) contains over 90 data fields and holds information such as chemical properties and descriptors, toxicity values, molecular and cellular interactions, and medical information. This information has been extracted from over 18,115 sources, which include other databases, government documents, books, and scientific literature. The focus of the T3DB is on providing mechanisms of toxicity and target proteins for each toxin. This dual nature of the T3DB, in which toxin and toxin target records are interactively linked in both directions, makes it unique from existing databases. It is also fully searchable and supports extensive text development, chemical structure, and relational query Database (HMDB) and DrugBank toxin/drug interaction prediction various fields. Overall, the various user and the advanced research'. A 'Please cite:' section lists: 'Lim E, Pon A, Djoumbou Y, Knox C 2010. Jan 30(Database Issue):D778'. The bottom part of the screenshot shows a detailed entry for Digoxin (T3D2670) with a table of properties:

Predicted LogP	2.3667
Route of Exposure	Injection or dermal contact. (W468)
Mechanism of Action	Digoxin binds to a site on the extracellular aspect of the alpha-subunit of the Na ⁺ /K ⁺ ATPase pump in the membranes of heart cells (myocytes) and decreases its function. This causes an increase in the level of sodium ions in the myocytes. This effect causes an increase in the length of the cardiac action potential, which when combined with the effects of digoxin on the parasympathetic nervous system, lead to a decrease in heart rate. Increased amounts of calcium are then stored in the sarcoplasmic reticulum and released by each action potential, which is unchanged by digoxin. This leads to increased contractility of the heart. Digoxin also increases vagal activity via its action on the central nervous system, thus decreasing the conduction of electrical impulses through the AV node. (S805)
Metabolism	Hepatic (but not dependent upon the cytochrome P-450 system). The end metabolites, which include 3-b-digoxigenin, 3-keto-digoxigenin, and their glucuronide and sulfate conjugates, are polar in nature and are postulated to be formed via hydrolysis, oxidation, and conjugation.
Toxicity Values	Not Available
Lethal Dose	Not Available
Carcinogenicity (IARC Classification)	Not Available
Uses/Sources	Digoxin is a plant toxin found in the foxglove plant (<i>Digitalis lanata</i>). It is used as a drug to treat various heart conditions, namely atrial fibrillation, atrial flutter and sometimes heart failure. (S805)
Minimum Risk Level	Not Available
Health Effects	Digoxin mainly affects the heart. (S805)
Symptoms	Adverse affects of digoxin include loss of appetite, nausea, vomiting, diarrhea, blurred vision, visual disturbances (yellow-green halos), confusion, drowsiness, dizziness, nightmares, agitation, and/or depression, as well as a higher acute sense of sensual activities. (S805)
Treatment	Treatment of dioxin overdose includes supportive measure and administration of the antidote, antidigoxin (DIGIBIND). Toxicity can also be treated with higher than normal doses of potassium. (S805) • S805 - Wikipedia. Digoxin. Last Updated 8 July 2009. • VIKR - Karjalanski J, Weinhouse F, Tonzar M, Gschik R, Varanamil and digoxin interactions in the rat

<http://www.t3db.ca>

- Comprehensive data on toxic compounds (drugs, pesticides, herbicides, endocrine disruptors, drugs, solvents, carcinogens, etc.)
- Detailed mechanisms, binding constants, target info, lots of ToxCast data
- >3600 toxic compounds
- >1900 reference spectra
- ~2100 toxic targets
- Supports sequence, spectral, structure, text searches as well as compound browsing
- Full data downloads



ContaminantDB



Welcome to ContaminantDB Version 1.0

The ContaminantDB is a unique bioinformatics resource that combines detailed contaminant data from different online references and databases on contaminants. The database currently houses 54,249 compounds. It is both modelled after and closely linked to the Human Metabome Database (HMDB) and DrugBank. The databases and sources used to gather contaminant data includes IARC Carcinogens Group 1, 2A, 2B, 3 and 4, Drugbank drugs and metabolites, Disinfection ByProducts, My Exposome Chemicals, ToxCast and Tox21 Chemicals, EPA High Production Volume Chemicals, OSHA Hazardous Chemicals, Clean Air Act Chemicals, T3DB Toxins, ECHA Substances of High Concern, DEA Chemicals, EPA Endocrine Screening Chemicals, EAFUS Chemicals, and OECD High Production Volume Chemicals.

CHEM003955	Epinephrine 51-43-4	$C_9H_{13}NO_3$ 183.204		<ul style="list-style-type: none">HMDB Contaminants - UrineSTOFF IDENT CompoundsToxCast & Tox21 Chemicals
CHEM003956	Carbachol chloride 51-83-2	$C_8H_{15}ClN_2O_2$ 182.649		<ul style="list-style-type: none">Clean Air Act ChemicalsHMDB Contaminants - UrineSTOFF IDENT CompoundsToxCast & Tox21 Chemicals
CHEM003957	Famphur 52-85-7	$C_{10}H_{16}NO_5PS_2$ 325.330		<ul style="list-style-type: none">Clean Air Act ChemicalsHPV EPA ChemicalsMy Exposome ChemicalsSTOFF IDENT CompoundsToxCast & Tox21 Chemicals
CHEM003958	2-Acetylaminofluorene 53-96-3	$C_{15}H_{13}NO$ 223.270		<ul style="list-style-type: none">Clean Air Act ChemicalsHPV EPA ChemicalsToxCast & Tox21 Chemicals

- Data on 54,249 probable or known chemical contaminants
- Expected to grow to 80,000+ by Sept. 2018
- Exp. MS data for 5000+ cmpds
- Pred. 54,000 EI-MS spectra, 150,000 ESI-MS/MS spectra
- Source or role information for most compounds
- >40% of the compounds in ContaminantDB are not found in PubChem or ChemSpider
- Supports spectral, structure and text searches as well as compound browsing

www.contaminantdb.ca



PhytoBank



The image shows two screenshots of the PhytoMap website. The top screenshot is the home page, featuring a navigation bar with 'PhytoMap', 'Viewer', 'Methods', 'Phyto-Compounds', and 'PhytoBank'. The main content area includes a 'Welcome to PhytoMap' heading and a description of the project's goal to create a comprehensive 'Google Map' of food crop metabolism. A photograph of several red tomatoes is displayed on the right. The bottom screenshot shows a 'Partial Listing of PhytoBank Metabolites' page. It includes a search bar and a table of metabolites.

PhytoBank ID CAS Number	Name	Structure	Formula Average Mass Mono Mass	Species
PMC000001	sarcostolide F		C ₂₀ H ₂₆ O ₄ 330.424 330.183109317	<i>Sarcophyton stolidotum</i>
PMC000002	Everminomicin C		C ₆₂ H ₈₂ Cl ₂ O ₃₃ 1436.29 1434.4897908	<i>Micromonospora carbonacea</i> Luedemann & Brodsky, 1965
PMC000003			C ₂₀ H ₁₈ NO ₄	<i>Citrus junos</i>

- 179,729 plant-derived compounds from more than 23,700 plant species including >8,318 food/crop plants and >2,439 medicinal plants
- >33% of the compounds in PhytoBank are not found in PubChem or ChemSpider
- Will offer same resources as HMDB, DrugBank, etc.

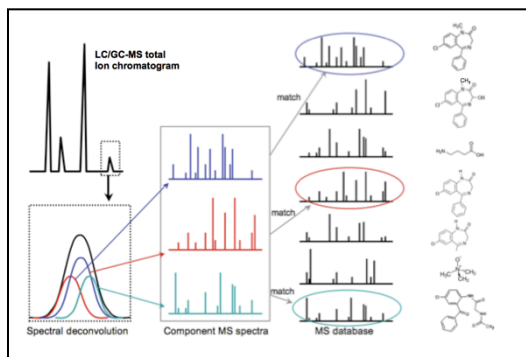
What Can We Predict?

In Silico Metabolomics

HMDB/DrugBank/T3DB

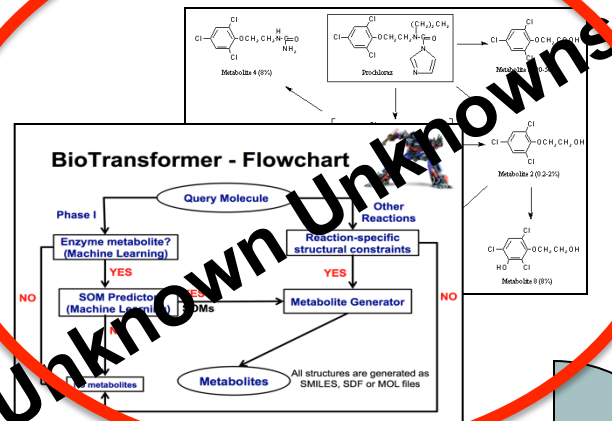


Known compounds (250,000)



Match observed spectra to predicted spectra to ID compounds

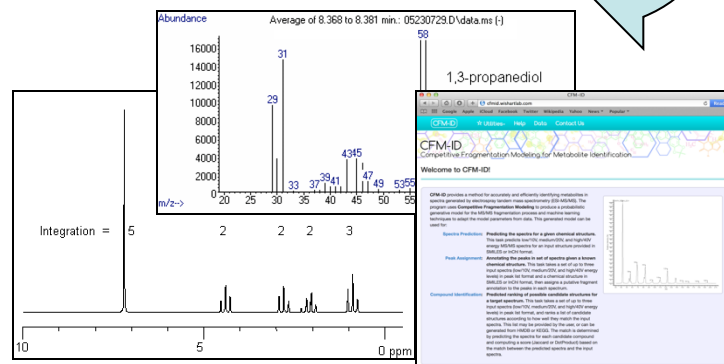
BioTransformer



Unknown Unknowns

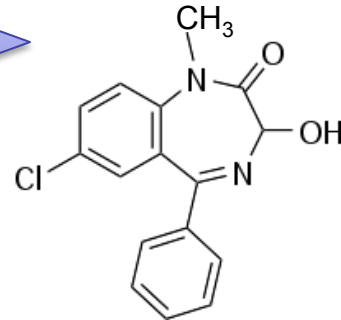
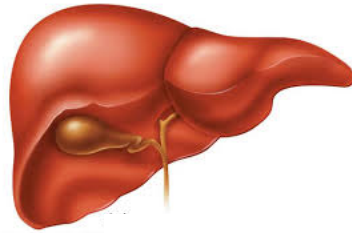
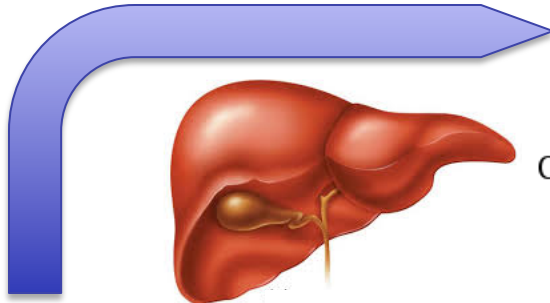
Predicted biotransformations (250,000 --> 5,000,000)

CFM-ID/NMRPred

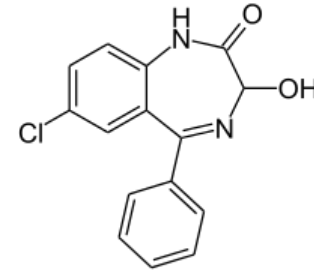


Predicted MS/MS, NMR, GC-MS Spectra of knowns + biotransformed

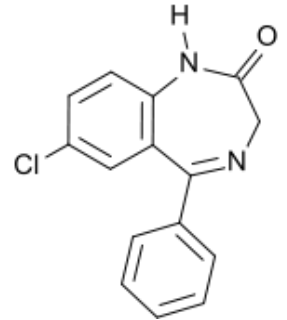
Examples of Biotransformation



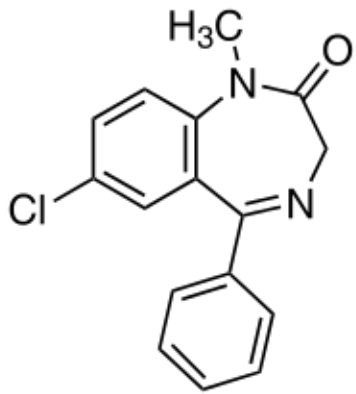
Tempazepam



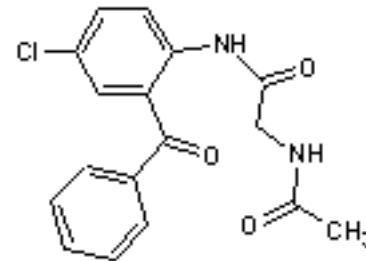
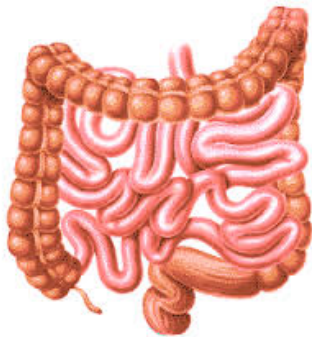
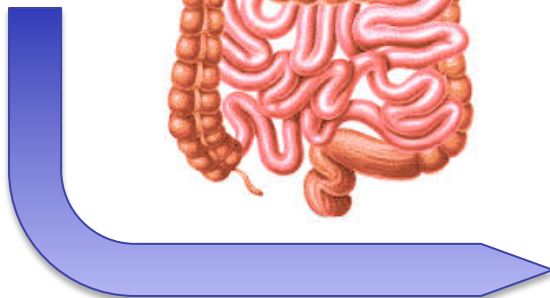
Oxazepam



Nordazepam



Diazepam



N-(2-Benzoyl-4-chlorophenyl)-2-acetamidoacetamide

Commercial Tools

IN SILICO PREDICTION OF METABOLISM: METEOR NEXUS (LHASA LTD)

Meteor Nexus then predicts the first generation metabolites by using:

- lipophilicity
- the presence or absence of structural features
- the relative likelihood of competing transformations

Choose	Structure	Name	Parent	Formula	Biotransformation Name	Phase	UGT	Exact Mass	Parent Mass Difference	LogP
<input type="checkbox"/>		MB9	MB8	C15H20O11	Glucuronidation of Aromatic Alcohols	Phase II	UGT	372.2693	176.0209	-1.92
<input type="checkbox"/>		MB10	MB8	C15H20O8	O-Sulphation of Aromatic Alcohols	Phase II	SULT	275.394	79.0962	-1.25
<input type="checkbox"/>		MB11	MB8	C15H20O8	O-Sulphation of Aromatic Alcohols	Phase II	SULT	275.394	79.0962	-1.25
<input type="checkbox"/>		MB5	Caffeic_a (Query)	C10H10O4	O-Methylation of Catechols	Phase II	COMT	194.0579	14.0153	1.42

Meteor-Nexus

ADMET Predictor(TM) C:\Program Files\Simulations Plus, Inc\ADMET Predictor\Demo2D.gmd

Generic	TPSA	TPSA_100	TPSA_1000	TPSA_10000	TPSA_100000	TPSA_1000000	ADMET_Risk	ADMET_Code
Acetamide	95.94	95.95	95.95	95.95	95.95	95.95	0.79	1.0
Acetic acid	93.36	93.36	93.36	93.36	93.36	93.36	2.25	1.0
Androstenedione	72.03	68.99	45.8	9.9	5.56	0.0		
Arabinose	78.2	78.2	78.2	78.2	78.2	78.2	1.63	2.0

ADMET Predictor

Metabolizer 5.2.1 - Untitled*

Metabolic Tree

- 0.13%
- 0.01%
- 27.20% Transmissibility
- 9.87% Accumulation
- 28.80% Stability
- 13.50% Production
- 0.13%

Metabolic Route

Alphahydroxy → Oxidation → Oxidation → Oxidation

Metabolizer

MetabolExpert

Work: default

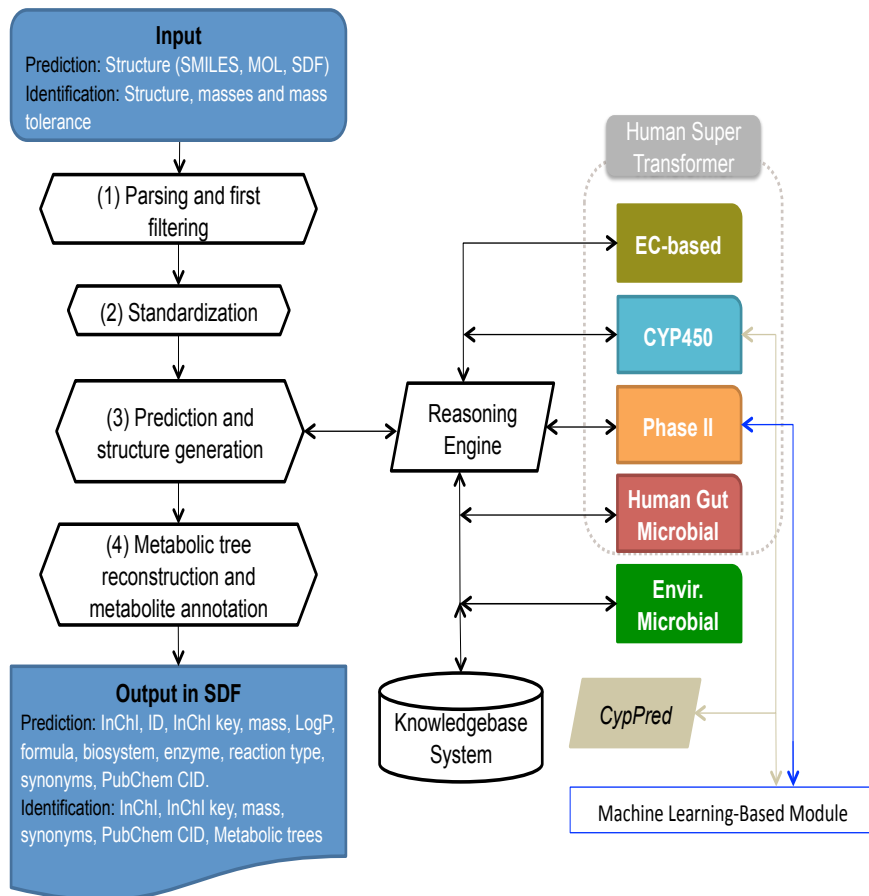
Result of AMPHETAMINE

View Options Report Erase Edit Get Info

Size: 100 Zoom: 30 (1287/351)

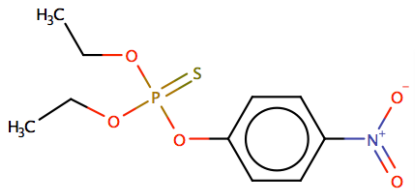
MetabolExpert

BioTransformer (Free)



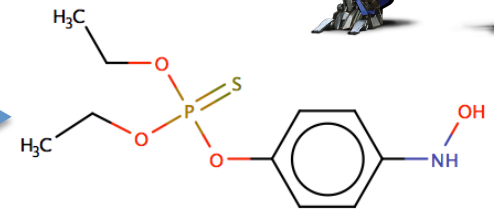
- A comprehensive tool for predicting metabolite structures that have been biotransformed through phase I, phase II, microbial, promiscuous and environmental processes
- Uses a large knowledgebase and a large set of heuristic (and machine-learned) biotransformation rules
- Performs much better than well-known commercial tools
- **Publicly released, manuscript submitted**

BioTransformer

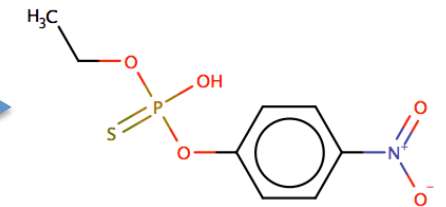


Parathion

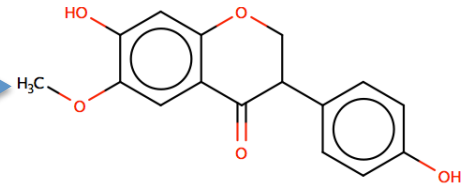
Phase I (163)



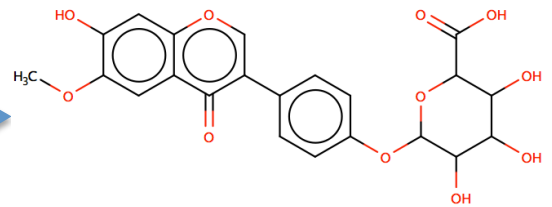
Env. Microbial (301)



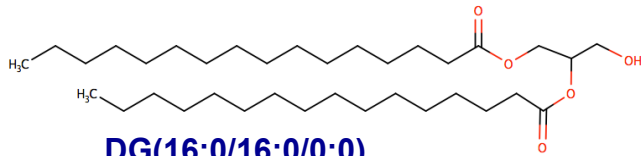
Human Gut (201)



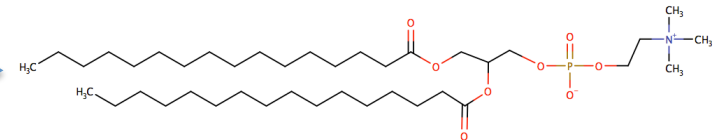
Phase II (74)



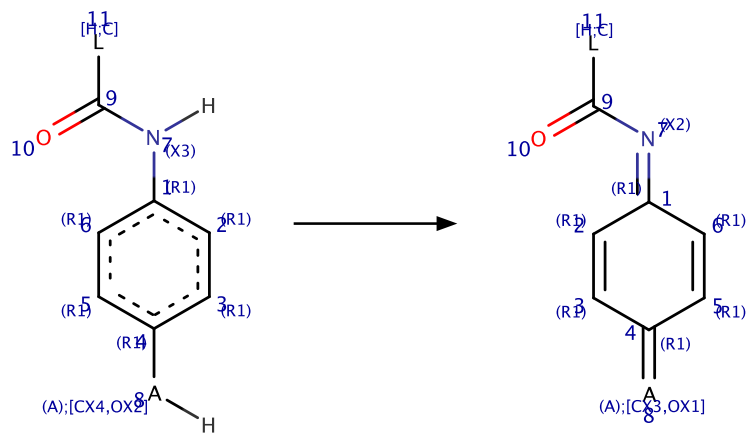
EC-based (408)



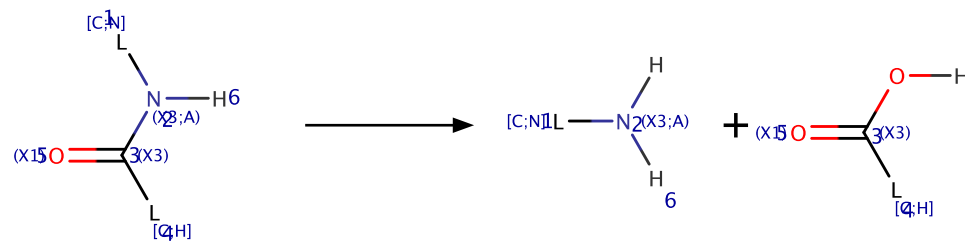
DG(16:0/16:0/0:0)



BioTransformer



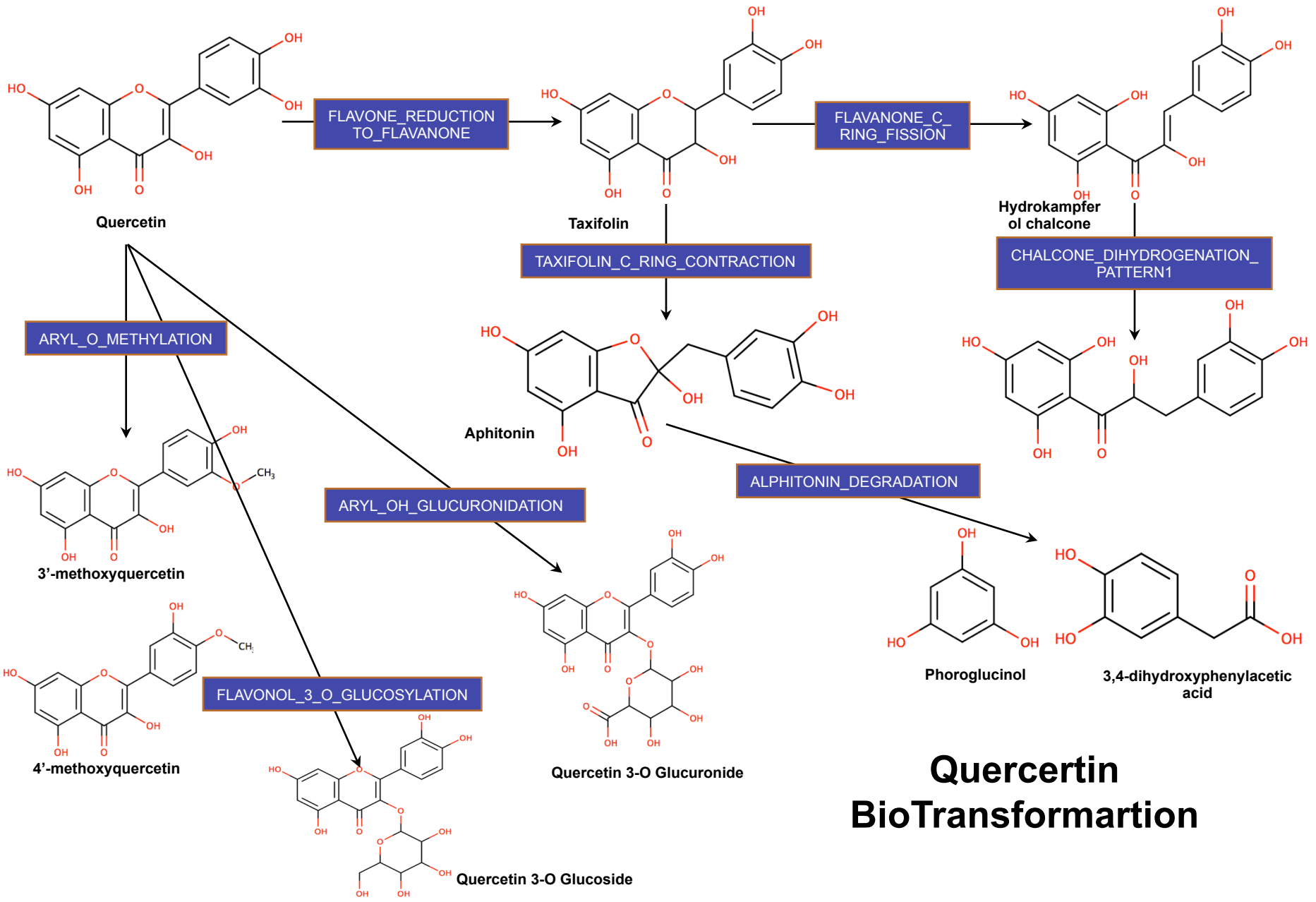
Oxidation of p-substituted anilides (BTMR1018)
Human (Liver) Phase I Metabolism



Hydrolysis of secondary amide (BTMR0704)
Environmental microbial (EAWAG-BBD)

	No. of enzymes	No. of biotransformation rules	No. of enzyme-rule associations	No. of covered biosystems
EC-based	285	408	459	2
CYP450	9	163	712	1
Human gut micro.	53	201	204	2
Phase II	9	74	81	2
Envir. microbial	1	301	301	1

Overall, 1,240 biotransformation rules were manually designed and tested. Overall, 2,150 reaction preference rules were implemented.



Quercetin BioTransformation

BioTransformer Evaluation



	BioTransformer	Meteor
True Positives	188	153
False Positives	198	281
False Negatives	35	70
Total No. of Predictions	386	431
Precision	0.49	0.35
Recall	0.84	0.69
# of reported metabolites	223	

1. Test set: 40 compounds (incl. drugs and pesticides)
2. Metabolism data was retrieved from >60 references
3. BioTransformer (v. 1.0.4) and Meteor Nexus (v. 3.0.1) were used for 1- step prediction of **human metabolism**

BioTransformer Evaluation



1. Test set: 60 compounds (incl. drugs, pesticides, food compounds, steroids)
2. Metabolism data was retrieved from 60+ references
3. BioTransformer (v. 1.0.4) and ADMET Predictor (v. 8.5.11) were used for 1- step prediction of **human CYP450 metabolism**

	BioTransformer	ADMET Predictor
True Positives	162	110
False Positives	188	122
False Negatives	18	70
Total No. Predictions	350	232
Precision	0.46	0.47
Recall	0.9	0.61
No. of Reported Metabolites	180	

1. Test set: 20 compounds (incl. endogenous metabolites, phytochemicals, and other xenobiotics)
2. Metabolism data was retrieved from >50 references
3. BioTransformer (v. 1.0.4) was used for 1-step prediction of **human and gut microbial metabolism**

	BioTransformer
True Positives	111
False Positives	49
False Negatives	17
Total No. Predictions	160
Precision	0.69
Recall	0.87
No. of Reported Metabolites	128

BioTransformer Updates



- **Added/updated:**
 - Available as a web server and a program, added metabolite ID option via m/z or molecular formula
- **Number of compounds expected**
 - 1-step reactions generate 5X as many compounds while 2-step reactions generate 20X as many compounds
 - Expect HMDB+BioTransformer will generate 2.2 million new compounds, all DBs+BioTransformer = 5 million cpds
- **Benchmark for computing time**
 - 1,000 FooDB compounds generate 5,071 human and gut microbial metabolites in 1 step (all enzymes)
 - 1h 29 mins (~5.35 s/compound per processor)

If you want compounds processed now, send your queries to
Yannick Djoumbou Feunang --- djoumbou@ualberta.ca

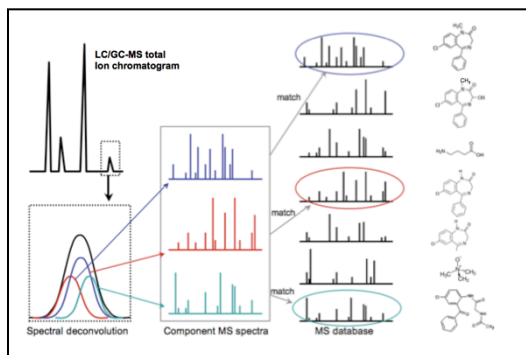
In Silico Metabolomics

HMDB/DrugBank/T3DB

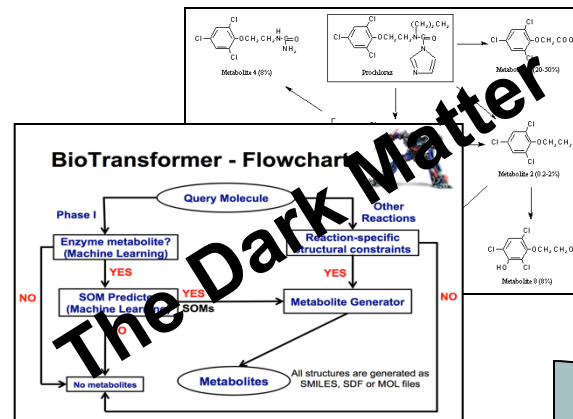
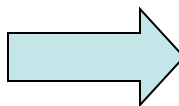
BioTransformer



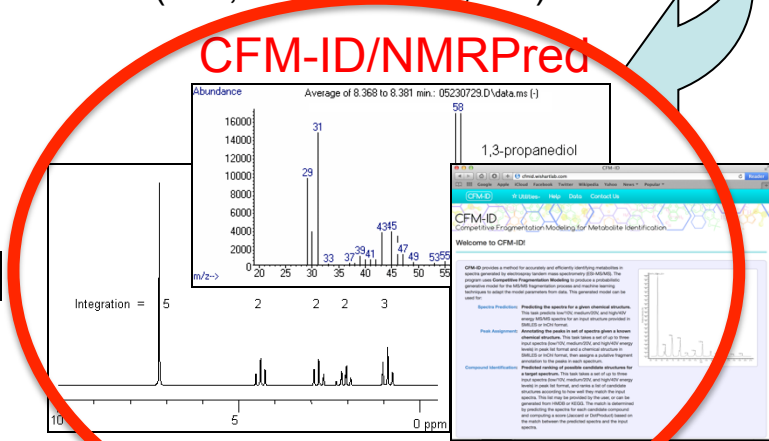
Known compounds (250,000)



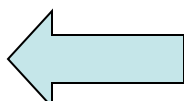
Match observed spectra to predicted spectra to ID compounds



Predicted biotransformations (250,000 --> 2,500,000)

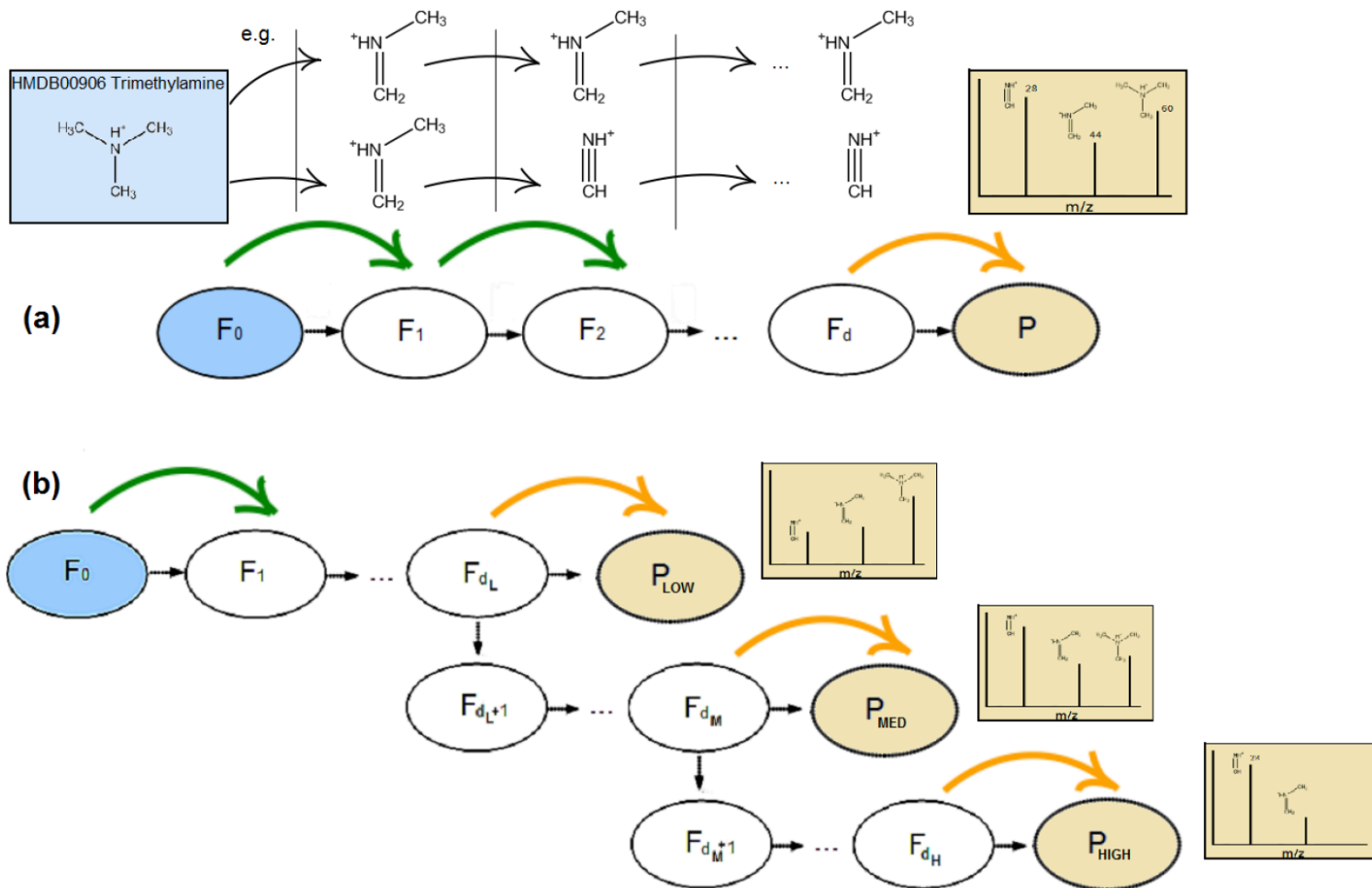


Predicted MS/MS, NMR, GC-MS Spectra of knowns + biotransformed



The Dark Matter

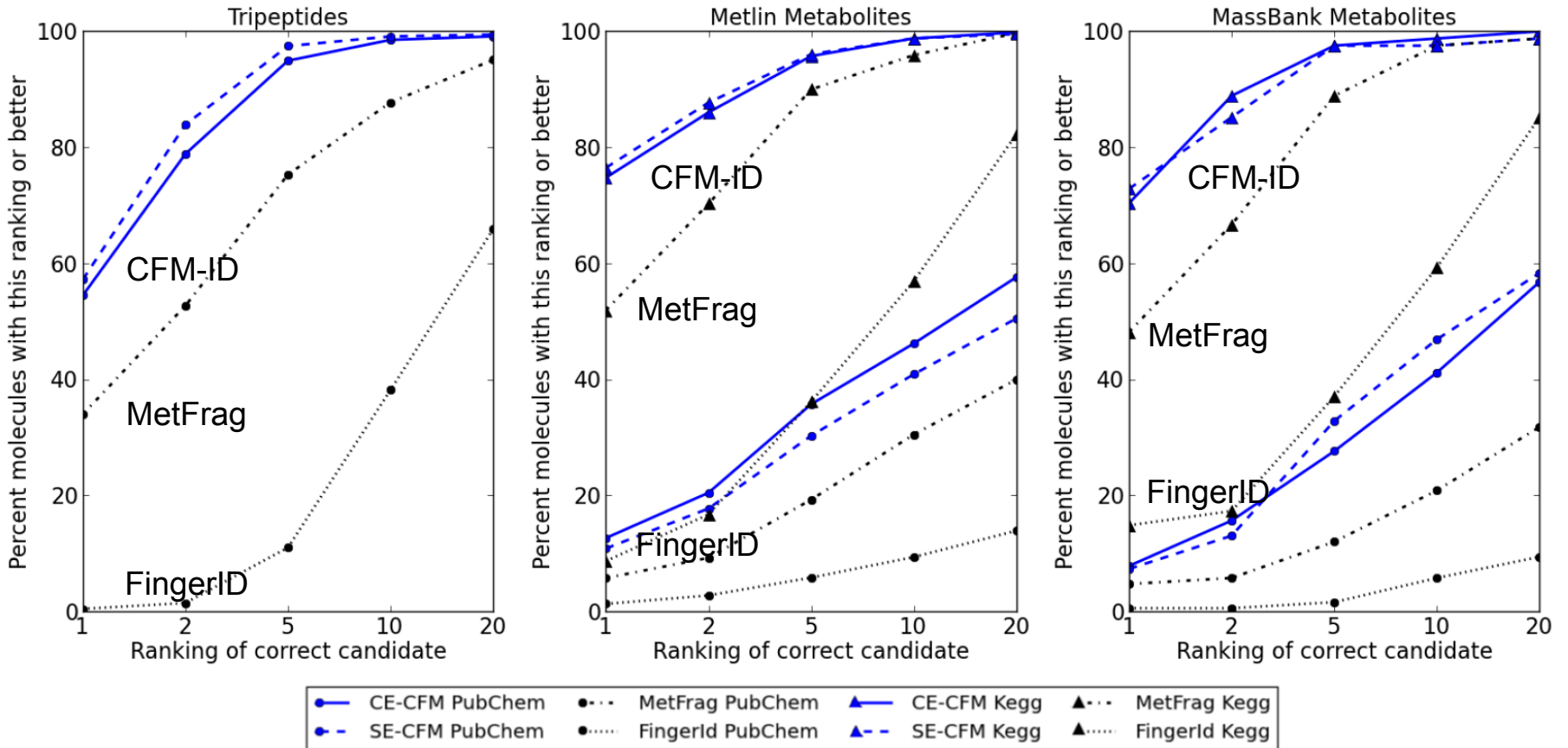
Competitive Fragment Modeling (CFM-ID)



CFM-ID

- Uses a large training set of high resolution MS/MS data of known compounds at low (10 eV) medium (20 eV) and high (40 eV) collision energies
- Employs an initially naïve chemical fragmenter that generates potential fragments and the corresponding MS/MS spectra
- The fragmenter slowly learns from its training data (via an HMM)
- *The more training data, the better the overall performance*

CFM-ID Performance



Performance

- **Significant performance improvement in CFM-ID and all other fragment or structure predictors if the database being searched is smaller or more targeted**
- **Significant improvement if multiple collision energies (10, 20, 40 eV) are used rather than a single collision energy**
- **80% correct for DB ~30,000 cmpds**
- **50% correct for DB ~1,000,000 cmpds**
- **20% correct for DB ~50,000,000 cmpds**

The CFM-ID Server

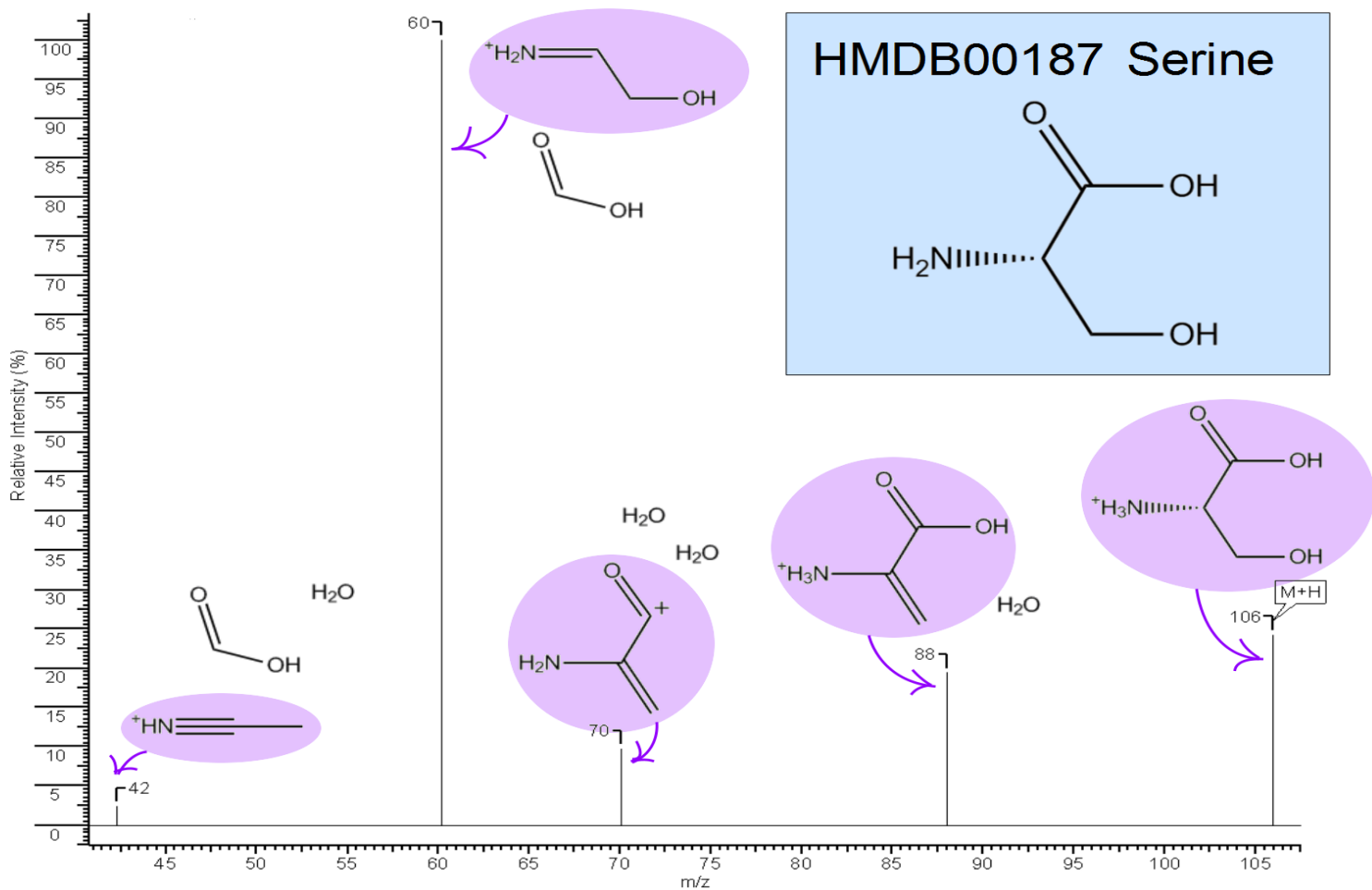
The screenshot displays the CFM-ID web server interface. The top navigation bar includes 'CFM-ID', 'Utilities', 'Help', 'Data', and 'Contact Us'. The main content area features a 'Welcome to CFM-ID!' message and a brief description of the server's purpose: 'CFM-ID provides a method for accurately and efficiently identifying metabolites in spectra generated by electrospray tandem mass spectrometry (ESI-MS/MS). The program uses Competitive Fragmentation Modeling to produce a probabilistic generative model for the MS/MS fragmentation process and machine learning techniques to adapt the model parameters from data. This generated model can be used for: Spectra Prediction, Peak Assignment, and Compound Identification.' Below this, there are three tabs for 'Low Energy Input MsMs Spectrum (10V), M+H', 'Medium Energy Input MsMs Spectrum (20V), M+H', and 'High Energy Input MsMs Spectrum (40V), M+H'. The 'Low Energy' tab is active, showing a mass spectrum plot with intensity on the y-axis (0 to 10K) and m/z on the x-axis (0 to 400). Below the plot, there are sections for 'Candidate Rankings' and a table with the following data:

Rank	Score	Structure	ID	Name	Chemical Formula	Mass	Compare
1	0.28571429		HMDB29101	Tyrosyl-Aspartate	C13H16N2O6	296.10083254	Current
2	0.19047619		HMDB28765	Aspartyl-Tyrosine	C13H16N2O6	296.10083254	Compare
3	0.076923077		HMDB11685	DHAP(B,0)	C11H21O7P	296.102489538	Compare

- A web server that predicts MS/MS spectra, annotates input MS/MS spectra and permits compound identification from input MS/MS spectra
- Matches predicted MS/MS spectra (from HMDB or KEGG) to input MS/MS spectra
- 1st and 2nd in the 2014 CASMI competition, used by winners of 2016 CASMI competition

<http://cfmid.wishartlab.com>

CFM-ID Example Output

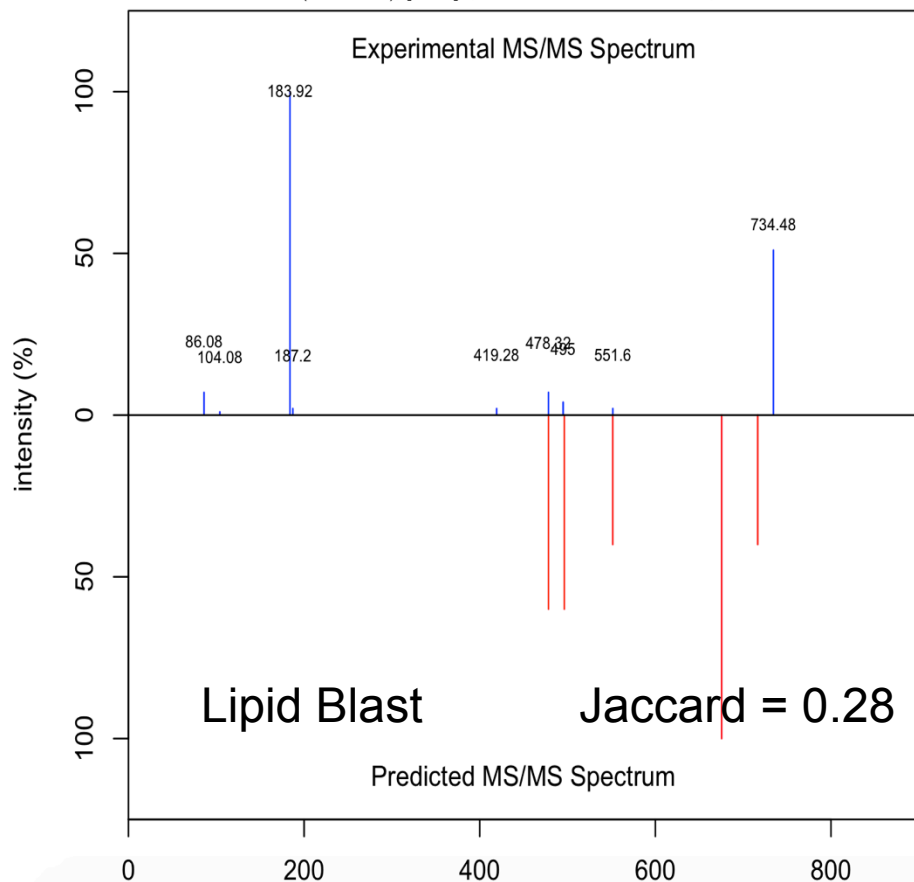


CFM-ID Updates

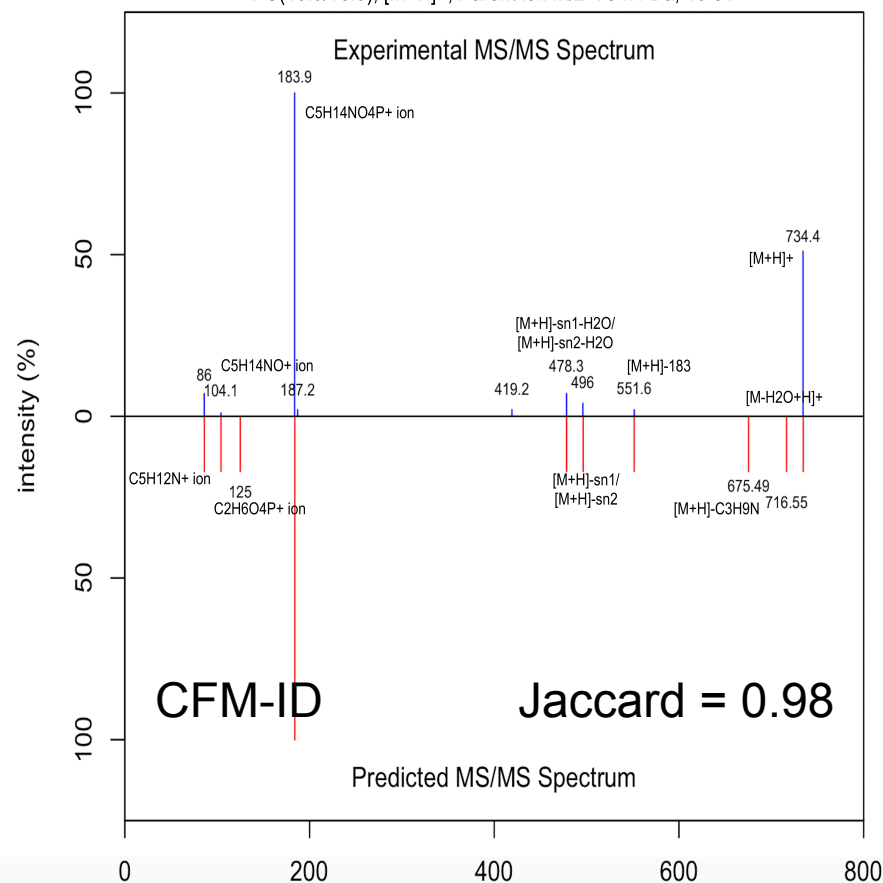
- **Version 3.0 completed (to be released in Dec):**
 - Significantly improved (5X) performance with respect to lipid MS/MS spectral prediction
 - Supports matches to known MS/MS spectra for better compound ID, improved scoring based on citations
- **Version 4.0 in progress (to be finished by 2019)**
 - Much larger training data set (4X larger) covering QTOF and OrbiTrap MS/MS spectra at multiple collision energies
 - Improved generative function, improved chemical and bond descriptors boosts spectral prediction performance by 30-40% over previous CFM-ID version
 - Combined improvements should increase overall performance by at least 50% (still not perfect)

CFM-ID 3.0 for Predicting 70,000 Lipid Spectra in the HMDB

PC(16:0/16:0); [M+H]⁺; Parent ion m/z = 734.4 Da; 40 eV



PC(16:0/16:0); [M+H]⁺; Parent ion m/z=734.4 Da; 40 eV



“Observables” Prediction

- **MS/MS spectral prediction alone will not be sufficient to ID all unknown compounds**
- **Other observables need to be included for confirmation such as RT (retention time), RI (retention index), CCS (collisional cross section), and gas phase IR or IR ion spectra**

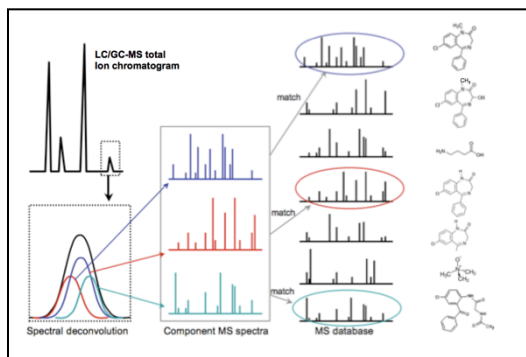
In Silico Metabolomics

HMDB/DrugBank/T3DB

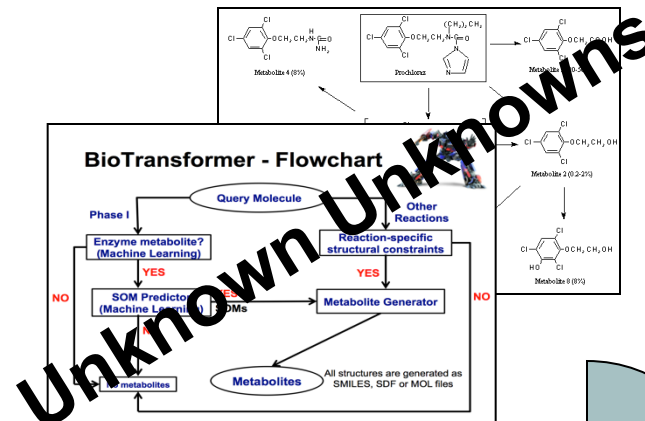
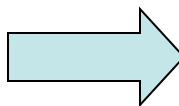
BioTransformer



Known compounds (250,000)

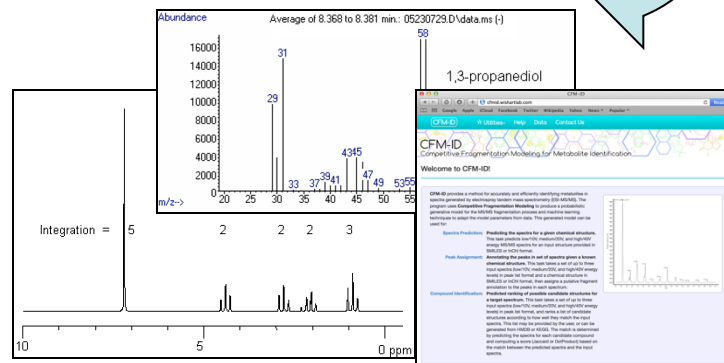


Match observed spectra to predicted spectra to ID compounds

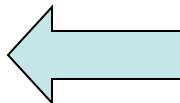


Predicted biotransformations (250,000 --> 5,000,000)

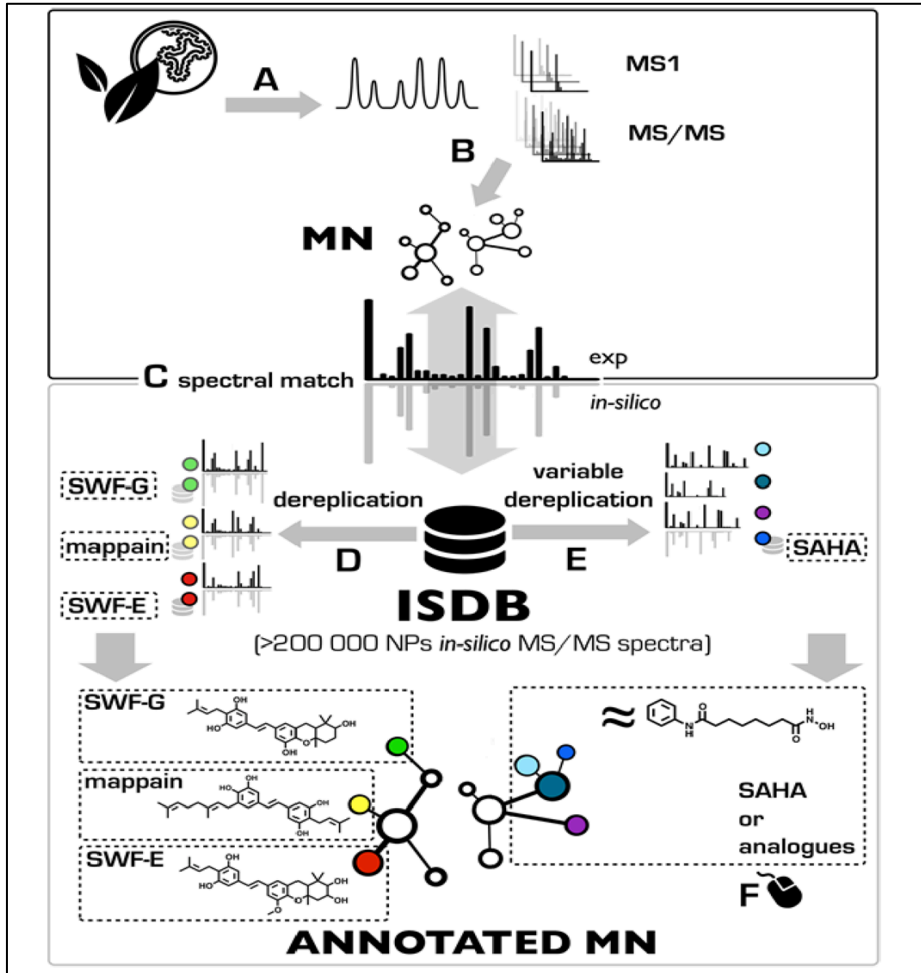
CFM-ID/NMRPred



Predicted MS/MS, NMR, GC-MS Spectra of knowns + biotransformed



Who Is Doing This?



PM Allard et al. *Anal. Chem.* 2016 88(6)
3317-3323

F Qiu et al. *Anal. Chem.* 2016 88(23)
11373-11383

Conclusions

- **Compound identification of unknown compounds by MS/MS analysis is hard**
- **We have insufficient MS/MS and compound resources now and the foreseeable future**
- ***In silico* methods offer a possible solution to the problem of inadequate spectral libraries and inadequate collections of compounds**
- **Predicted compound libraries and predicted MS/MS spectra are still imperfect, but they are getting better every year**
- **These *in silico* methods are already being used by several groups in natural product analysis**

Thanks To...

- Yannick Djoumbou Feunang
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- Tanvir Sajed
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- Zinat Sayeeda
- Elvis Lo
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- Augustin Scalbert
- Sandeep Singhal
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- Jason Grant
- Arnau Serra-Cayuela
- Yifeng Liu
- Rupasri Mandal
- Vaness Neveu
- Allison Pon
- Craig Knox
- Mike Wilson
- Claudine Manach



CIHR IRSC



GenomeAlberta



GenomeCanada



Canada Foundation for Innovation
Fondation canadienne pour l'innovation



THE METABOLOMICS
INNOVATION CENTRE



NSERC
CRSNG