

# Repository-scale Co- and Re-analysis of Tandem Mass Spectrometry Data

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## 52 **Author Contributions**

53 AKJ, MW, and PD developed the ReDU concept.

54 AKJ, MW, and CMA wrote code and engineered the ReDU infrastructure.

55 AKJ, CMA, RSA, SA, AAA, GA, AA, AB, SB, AB, AMCR, RC, EOE, JJJvdH, JMG, ECG, MH, KJ, ZK, ALG, AL,  
56 LIM, KLM, MJM, AVM, RCM, YAM, NHN, LFN, ME, MNE, MP, DP, RQ, NS, FV, AV, and KW curated  
57 metadata enabling ReDU.

58 AKJ, MW, CMA, SAJ, LM, ME, JJJvdH, JMG, MP and PCD tested the ReDU infrastructure and provided  
59 feedback.

60 AKJ, MW, CMA, ME, JJJvdH, RK, NB, and PCD wrote and edited the manuscript.

61 RK, NB, and PCD provided supervision and funding support.

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## 63 **Ethics Declaration**

64 Pieter C. Dorrestein is a scientific advisor for Sirenas LLC.

65 Mingxun Wang is a consultant for Sirenas LLC and the founder of Ometa labs LLC.

66 Alexander Aksenov is a consultant for Ometa labs LLC.

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## 68 **Abstract**

69 Metabolomics data are difficult to find and reuse, even in public repositories. We, therefore, developed the  
70 Reanalysis of Data User (ReDU) interface (<https://redu.ucsd.edu/>), a community- and data-driven approach that  
71 solves this problem at the repository scale. ReDU enables public data discovery and co- or re-analysis via  
72 uniformly formatted, publicly available MS/MS data and metadata in the Global Natural Product Social Molecular  
73 Networking Platform (GNPS), consistent with findable, accessible, interoperable, and reusable (FAIR)  
74 principles.<sup>1</sup>

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## 76 **Results and Discussion**

77 Many simple but important questions can be asked using repository-scale public data. For example, what  
78 human biospecimen or sampling location is best for detecting a given drug? Or what molecules are found in  
79 humans <2 years old? Current metabolomics repositories typically require manual navigation and conversion of  
80 thousands of different vendor-formatted files with inconsistent metadata formats, and developing data integration  
81 algorithms, greatly complicating analyses.

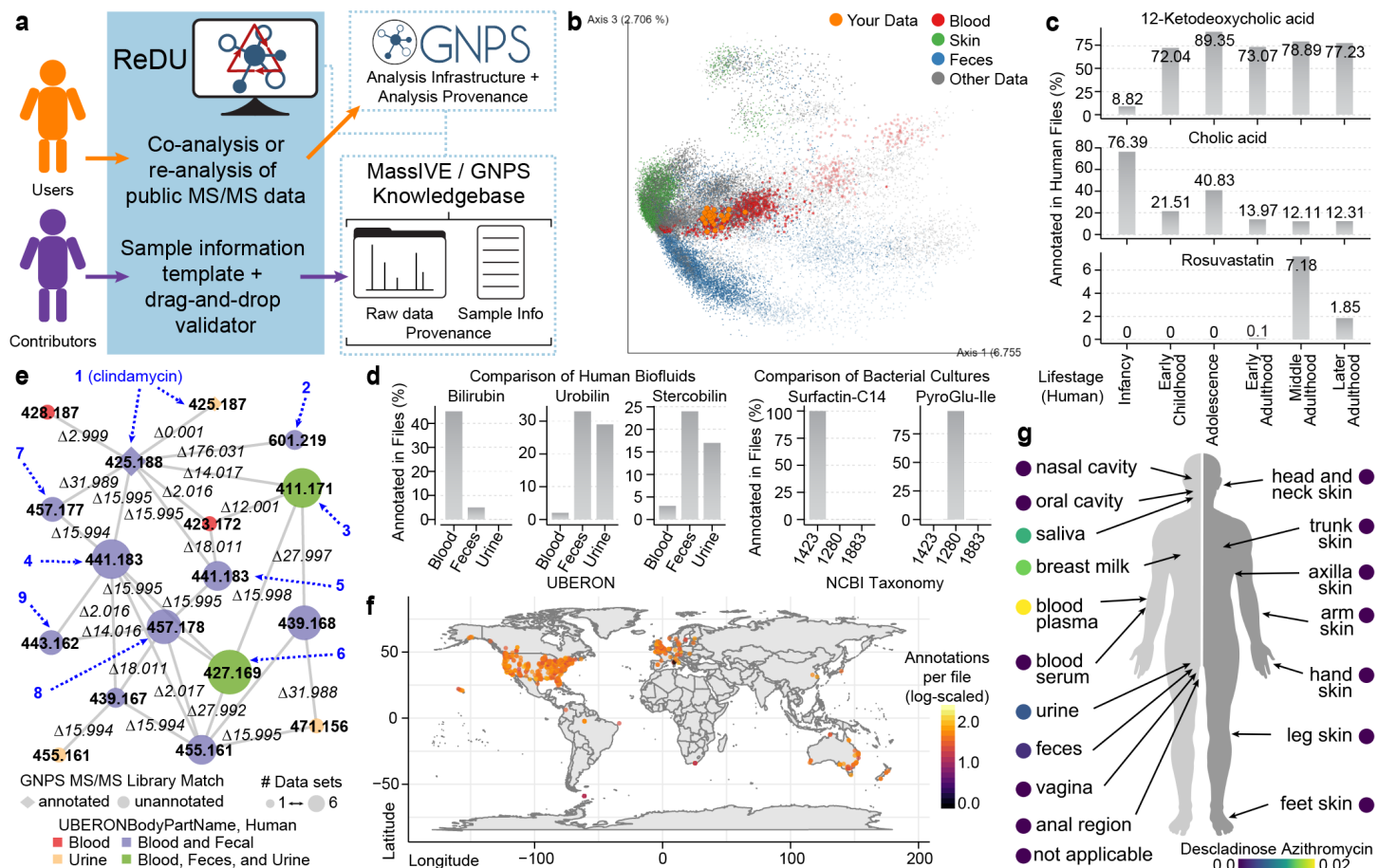
82 ReDU addresses FAIR principles by enabling users to find and choose files (**Fig 1a**). This is possible  
83 because ReDU formats sample information consistently via a template and drag-and-drop validator backed by  
84 standard controlled vocabularies and ontologies (e.g. NCBI taxonomy,<sup>2</sup> UBERON<sup>3</sup>, Disease Ontology<sup>4</sup> and MS  
85 ontology), and includes geographical location (important for natural products and environmental samples). ReDU  
86 automatically uses all public data in the GNPS/MassIVE repository that has the corresponding ReDU-compliant  
87 sample information. 34,087 files in GNPS are ReDU-compatible including natural and human-built environments,  
88 human and animal tissues, biofluids, food, and other data from around the world (**Fig 1f**), analyzed using different  
89 instruments, ionization methods, sample preparation methods, etc. From the 103,230,404 million MS/MS spectra  
90 included in ReDU, 4,528,624 spectra were annotated (rate of 4.39% with settings yielding ~1% FDR) as one of  
91 13,217 unique chemicals (**Table S1**).<sup>5,6,7</sup>

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93 Uniformity of data and sample information in ReDU enables metadata-based and repository-scale  
94 analyses (**Fig. 1b-g**). Chemical explorer enables selection of a molecule and retrieval of its associations with the  
95 metadata, *i.e.* sample information association. For instance, selecting 12-ketodeoxycholic acid (filtering to  
96 include human feces) revealed it was observed after infancy (**Fig 1c**), whereas cholic acid displayed the opposite  
trend, coupled to the developing microbiome. Similarly, rosuvastatin was found in adults matching prescription

demographics. Another approach enabled is chemical enrichment analysis. For example, human blood, feces, and urine differed by bilirubin, urobilin, and stercobilin (**Fig 1d**). Bilirubin was more frequently annotated in blood, and urobilin and stercobilin were most often annotated in feces.<sup>8</sup> Similarly, comparison of bacterial cultures revealed differences in annotation of surfactin-C14 (observed in *Bacillus subtilis*) and cholic acid (observed in *Streptomyces*). ReDU enables reanalysis based on metadata-selected files for molecular networking.<sup>5,9,10</sup> Re-analyzing human blood plasma and serum, urine, and fecal samples, networked 5,053,666 MS/MS spectra (~5.6% annotated) and included annotations to clindamycin. Clindamycin's (1) molecular family matched multiple datasets and sample types (**Fig 1e**). Using propagation through molecular networking (e.g. delta mass and MS/MS spectral interpretation), we annotated clindamycin metabolites (2-9) **Fig S1-S2, Table S2-S3**. *N*-desmethylclindamycin sulfoxide (6) was observed in multiple sample types across six different datasets. At the repository scale, we can map sample geographical location and identify the number of chemical annotations for each sample (**Fig 1f**), or locate specific molecules of interest (e.g. drugs) by mapping on the human body offline (**Fig 1g, Video S1**). These are representative analyses uniquely enabled by the ReDU infrastructure.

ReDU makes public MS/MS data FAIR and connects MassIVE (a data repository recommended by Nature publishing journal for metabolomics and proteomics data) to GNPS (an analysis environment), thereby integrating public data deposition, sample information curation, and data analysis. ReDU's utility will continue to grow as more data is uploaded to GNPS/MassIVE, and public reference libraries expand, making ReDU a resource developed by the community and FAIR for the community.



**Fig 1. ReDU workflow and illustrative applications.** (a) The ReDU framework. (b) Users can co-analyze their data via projection onto public data visualized using EMPEROR<sup>11</sup>. PCA was performed on GNPS annotations (level 2/3).<sup>7</sup> Human blood plasma samples (orange) from rheumatoid arthritis patients. Sample points size and color were set using UBERON ontology and opacity was set using NCBI taxonomy filtering (your data, 1.0; 9606|Homo sapiens, 0.7; and all other data, 0.25). (c) Metadata filters were used to select human fecal samples (NCBI taxonomy, 9606|Homo sapiens; UBERON, feces) and launch sample information enrichment. Chemical explorer

123 was used to select 12-Ketodeoxycholic acid, cholic acid, and rosuvastatin. Lifestages: Infancy (<2 yrs), n=1859;  
124 Early Childhood (2 yrs < x ≤ 8 yrs), n=93; Adolescence (8 yrs < x ≤ 18 yrs), n=169; Early Adulthood (18 yrs < x  
125 ≤ 45 yrs), n=995; Middle Adulthood (45 yrs < x ≤ 65 yrs), n=933; and Later Adulthood (> 65 yrs), n=325. **(d)**  
126 Metadata filters were used to select human blood, feces, and urine into different groups and launch chemical  
127 enrichment analysis. Bilirubin, urobilin, and stercobilin are illustrative of the chemical differences between the  
128 groups. Similarly, bacterial cultures of 1423|*Bacillus subtilis* (n=89), 1280|*Staphylococcus aureus* (n=49), and  
129 1883|*Streptomyces* (n=7) were selected into groups using filters. Surfactin-C14 and PyroGlu-Ile were two  
130 exemplar chemicals observed to be different between groups. **(e)** Human blood, feces, and urine were selected  
131 using ReDU metadata filters and re-analyzed together using molecular networking. A portion of the molecular  
132 network associated with clindamycin is displayed. Nodes are colored by the UBERON ontology in which it was  
133 detected: node size reflects the number of MassIVE datasets in which it was detected, node shape indicates  
134 whether annotated via library search (annotated, diamond or unannotated, circle). Putatively annotated  
135 clindamycin metabolites are indicated using dashed arrows and numbers, blue, corresponding to the proposed  
136 structures **(1-9)**. **(f)** ReDU sample locations (includes public data from samples of environmental, natural  
137 products and other cohorts for which this information is provided) colored by number of annotations per file  
138 (latitude and longitude), log<sub>10</sub>-scale, using the ReDU database. **(g)** ReDU database (filtered to include human  
139 data) analyzed to visualize the distribution of chemical annotations tagged as drug or drug metabolite using 'ili'.<sup>12</sup>  
140 Descladinose azithromycin was detected in blood plasma, breast milk, saliva, urine, and fecal samples. The  
141 number of annotations are divided by the number of files per sample type (*i.e.* UBERON). A distribution map of  
142 all drugs in ReDU is provided.

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## 152 Data Availability

153 All curated sample information can be downloaded from the ReDU homepage (<https://redu.ucsd.edu/>) by  
154 selecting "Download Database." The current version of the ReDU information is archived in the GNPS/MassIVE  
155 ([gnps.ucsd.edu](https://gnps.ucsd.edu)) repository. The accession number is MSV000084206.

## 156 Code Availability

157 All code for ReDU is available in GitHub (<https://github.com/mwang87/ReDU-MS2-GNPS>) with corresponding  
158 documentation (<https://github.com/mwang87/ReDU-MS2-Documentation>).

## 159 References

- 160 1. Wilkinson, M. D. et al. *Sci. Data*. **3**, 160018 (2016).
- 161 2. Federhen, S. *Nucleic Acids Res.* **40**, D136–D143 (2012).
- 162 3. Mungall, C. J., Torniai, C., Gkoutos, G. V., Lewis, S. E. & Haendel, M. A. *Genome Biol.* **13**, R5 (2012).
- 163 4. Schriml, L. M. & Mitraka, E. *Mammalian Genome*. **26**, 584–589 (2015).

- 164 5. Wang, M. et al. *Nat. Biotechnol.* **34**, 828–837 (2016).  
165 6. Scheubert, K. et al. *Nat. Commun.* **8**, 1494 (2017).  
166 7. Sumner, L. W. et al. *Metabolomics*. **3**, 211–221 (2007).  
167 8. Pelley, J. W. Elsevier’s Integrated Review Biochemistry. Mosby (2012). doi:10.1016/B978-0-323-07446-  
168 9.00009-X  
169 9. Quinn, R. A. et al. *Trends in Pharmacological Sciences*. **38**, 143–154 (2017).  
170 10. Aron, A. T. et al. *ChemRxiv* (2019). doi:10.26434/CHEMRXIV.9333212.V1  
171 11. Vázquez-Baeza, Y., Pirrung, M., Gonzalez, A. & Knight, R. *Gigascience*. **2**, 1–4 (2013).  
172 12. Protsyuk, I. et al. *Nat. Protoc.* **13**, 134–154 (2018).