

1 The SIRIUS API: Programmatic Access to Metabolite 2 Annotation

3 **Jonas A. Emmert** ^{1,2}, **Martin A. Hoffmann**  ³, **Marcus Ludwig**  ³, **Kai**
4 **Dührkop**  ³, **Martin Engler-Lukajewski**  ³, **Lukas Scholz**  ¹, **Nils A.**
5 **Haupt**  ¹, **Sebastian Böcker**  ^{1,¶}, and **Markus Fleischauer**  ^{1,3¶}

6 ¹ Chair for Bioinformatics, Institute for Computer Science, Friedrich Schiller University Jena, Jena,
7 Germany  ² International Max Planck Research School "Chemical Communication in Ecological
8 Systems", Max Planck Institute for Chemical Ecology, Jena, Germany  ³ Bright Giant GmbH, Jena,
9 Germany ¶ Corresponding author

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

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Submitted: 01 January 1970

Published: unpublished

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10 Summary

11 Metabolomics plays a crucial role in understanding biological systems, with mass spectrometry
12 being the method of choice for analyzing complex metabolite mixtures, due to its high sensitivity
13 and throughput. However, compound annotation remains a major bottleneck in metabolomics
14 workflows, as for the majority of detected signals no reference spectra are found in spectral
15 libraries. This co-called "dark metabolome" ([da Silva et al., 2015](#)) necessitates approaches that
16 can annotate metabolites without relying on spectral library matching. SIRIUS ([Böcker et al.,](#)
17 [2009; Dührkop et al., 2019](#)) is a comprehensive software framework for annotating metabolites
18 and other small molecules from tandem mass spectrometry (MS/MS) data without the need of
19 reference spectra, while still supporting spectral matching when reference spectra are available.

20 Here, we introduce a RESTful API for SIRIUS that adheres to the OpenAPI Specification. This
21 enables programmatic access and seamless integration of SIRIUS into computational workflows
22 while future-proofing projects towards later changes by separating the storage layer from the
23 access layer. Built on this foundation, we present *PySirius* and *RSirius*, client libraries for
24 Python and R available via conda-forge, alongside a SIRIUS Java SDK. These clients facilitate
25 integration into bioinformatics pipelines, statistical analyses, data visualization workflows,
26 and other software. Actively maintained on GitHub, the API and client libraries democratize
27 access to state-of-the-art metabolite annotation and enable reproducible research practices in
28 computational metabolomics.

29 Statement of need

30 Many specialized metabolomics tools, including SIRIUS for annotation of MS/MS spectra with
31 molecular formulas, structures, and compound classes, are designed as standalone applications
32 accessed through command-line or graphical interfaces. Clearly, a graphical user interface is
33 of massive value for allowing inexperienced user accessing a computational tool ([Ludwig et](#)
34 [al., 2020](#)). Yet, this architecture also creates barriers for the integration into automated
35 computational workflows and advanced data mining: Users must parse text outputs or
36 intermediate files, and cannot access granular results programmatically.

37 This challenge was exemplified in SIRIUS 5, where a project directory of plain-text files served
38 as both the storage layer and the primary point of user access. This tight coupling created
39 a difficult dilemma: either break established user workflows by changing file formats to add
40 new features, or hinder innovation to maintain backward compatibility. The transition to an

41 embedded database in SIRIUS 6 solved the innovation bottleneck, but by removing direct file
42 access, it increases the barrier for users needing to programmatically integrate SIRIUS results.

43 The SIRIUS API addresses these limitations by providing programmatic access to all SIRIUS
44 capabilities through a standardized, storage-independent interface. This decoupled architecture
45 separates the data storage layer from the data access layer. It enables direct retrieval of
46 specific information at any granularity, and allows true integration into existing data analysis
47 environments while simultaneously permitting the internal storage formats to evolve without
48 breaking user-facing compatibility.

49 Furthermore, the SIRIUS API enables a more interactive data analysis paradigm. Users
50 are no longer required to load entire project spaces (which can be gigabytes in size) into
51 memory. Instead, they can fetch specific, fine-grained results on demand or selectively perform
52 (re-)computations with new parameters.

53 State of the field

54 Python and R dominate metabolomic data analysis, backed by libraries like *pyOpenMS* (Röst
55 et al., 2014), *matchms* (Huber et al., 2020), and the RforMassSpectrometry initiative including
56 *Spectra* and *Metabonaut* (Louail & Rainer, 2025). *PySirius* and *RSirius* bring SIRIUS into these
57 established ecosystems, enabling seamless workflows alongside statistical analyses, visualizations,
58 and machine learning models.

59 SIRIUS integrations for multiple widely used mass spectrometry tools such as *MZmine* (v4.8)
60 (Schmid et al., 2023), *Agilent MassHunter Explorer* (v2.0) and *RuSirius* (RforMassSpectrometry)
61 are already based on the SIRIUS API, further demonstrating community demand for
62 programmatic access to SIRIUS.

63 Software Design

64 The SIRIUS Application Programming Interface (API), initially released with SIRIUS version
65 6.0.0, is a major architectural update that offers several strategic advantages:

66 Comprehensive programmatic access to functionality

67 The API exposes the complete spectrum of SIRIUS analytical capabilities through programmatic
68 interfaces, encompassing spectral library searching, computation of fragmentation trees (Böcker
69 & Dührkop, 2016; Böcker & Rasche, 2008), annotation of molecular formulas (Dührkop et
70 al., 2013) (orange in Figure 1), structure database search with CSI:FingerID (Dührkop et al.,
71 2015), compound class prediction with CANOPUS (Dührkop et al., 2021), *de novo* molecular
72 structure prediction with MSNovelist (Stravs et al., 2021) and additional methods. This avoids
73 the need to parse command-line interface outputs or manipulate intermediate file formats.

74 Fine-grained and direct data retrieval

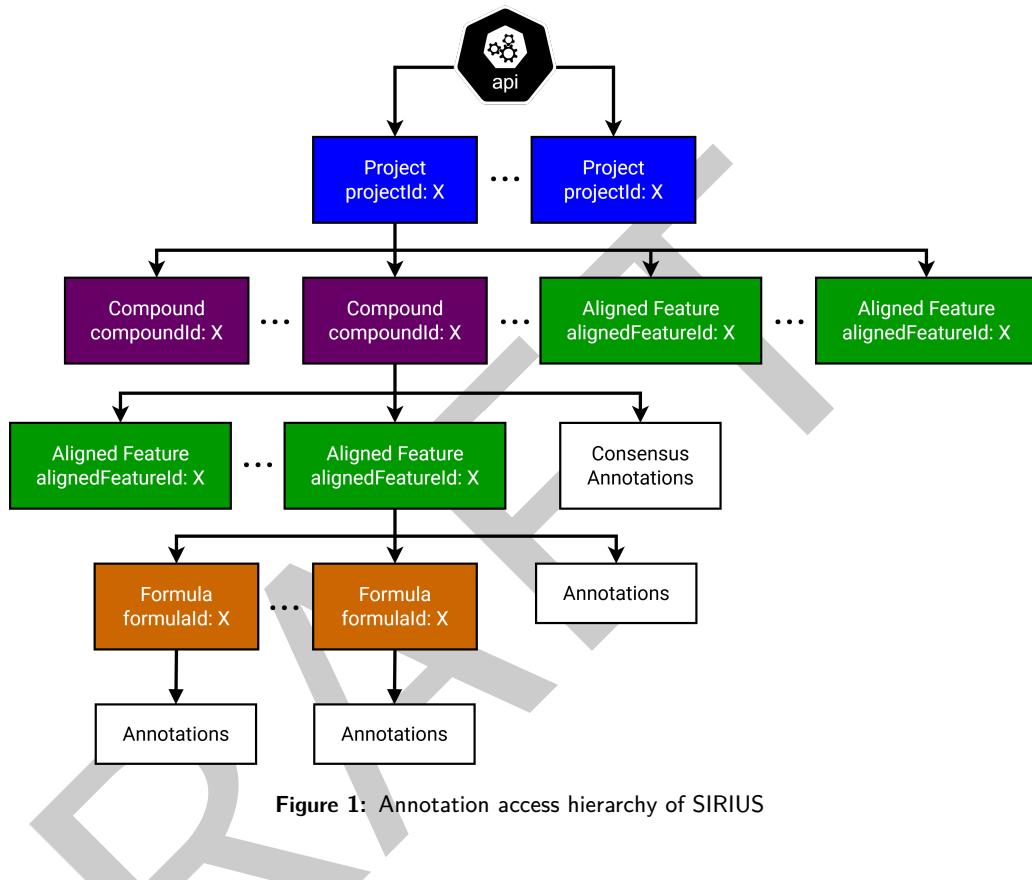
75 The API allows direct retrieval of specific data elements at arbitrary levels of granularity
76 (Figure 1). This spans from individual feature-level results to project-wide summaries, thereby
77 enabling more efficient and precisely targeted analytical workflows.

78 Decoupled architectural design

79 The API separates the data persistence layer from the data access layer, which permits
80 modifications to core SIRIUS functionality without invalidating existing project data structures.
81 This architectural decision ensures forward compatibility for future SIRIUS updates, addressing
82 the compatibility limitations encountered during the transition from SIRIUS version 5 to 6.

83 **Platform-agnostic integration capabilities**

84 The implementation according to the OpenAPI Specification guarantees that SIRIUS
85 functionality remains accessible from any programming language ecosystem equipped with
86 standard HTTP communication libraries.



106 mass spectrometry data from various formats. It further provides access to prediction vector
107 definitions utilized by CSI:FingerID and CANOPUS.

108 **Compounds API**

109 The compounds API module manages compound entities (purple in [Figure 1](#)) within projects.
110 Compound manipulation operations encompass addition, deletion, and information retrieval.

111 **Features API**

112 The features API module facilitates manipulation of aligned features (green in [Figure 1](#)) as well
113 as retrieval of their associated annotation results. Analogous to the compounds API, available
114 operations include addition, deletion, and information retrieval. Accessible annotations include
115 essentially all computational results generated by methods implemented within SIRIUS.

116 **Searchable Databases API**

117 The searchable databases API module manages custom spectral and structural database
118 resources. This module supports creation and deletion of custom databases, import of new
119 entries into existing databases, retrieval of database metadata, and enumeration of all custom
120 and built-in databases available within the SIRIUS environment. As spectra must be annotated
121 with a structure, any spectral library also functions as a structure database.

122 **Jobs API**

123 The jobs API module controls annotation job workflows. Users can define custom job
124 configurations specifying which analytical methods to apply to designated datasets. These
125 jobs can then be executed, monitored, terminated, or deleted as needed. Job configurations
126 can be saved for reuse, and their command-line equivalents retrieved for reproducibility.

127 **Client libraries**

128 Built on the RESTful API, *PySirius*, *RSirius*, and a Java client library provide idiomatic
129 interfaces for Python, R, and Java users, respectively. The libraries are automatically generated
130 from the OpenAPI specification using the [OpenAPI Generator](#), occasionally requiring custom
131 templates and automated patches for full compatibility. The SIRIUS development team actively
132 maintains all client libraries to ensure continued compatibility and feature parity with the API.

133 All client libraries include *SiriusSDK* helper classes that manage the SIRIUS REST service
134 lifecycle. These classes enable starting the service from within code with automated process
135 detection, shutdown and restart capabilities, and attachment to existing instances. Methods
136 return a central API client for streamlined access to all API modules and endpoints. Researchers
137 can now analyze tandem mass spectrometry data with SIRIUS using standard Python, R and
138 Java code. This includes loading data, running jobs, and retrieving results without leaving
139 their analytical environment. The clients enable sharing of complete workflows in contained
140 environments rather than describing multi-tool methodologies.

141 **Research Impact Statement**

142 SIRIUS and its contained methods have been named “methods to watch” in *Nature Methods*
143 on multiple occasions ([Singh, 2020, 2023](#)). The core methods, such as CSI:FingerID and
144 CANOPUS, are established standards in the field, evidenced by hundreds of citations for
145 their foundational papers. Furthermore, the web services for these tools recently surpassed a
146 combined [one billion queries](#), demonstrating large scale adoption.

147 The Java client has enabled integration of SIRIUS into major metabolomics platforms. *MZmine*
148 has incorporated SIRIUS functionality directly into its workflow since version 4.8, while

149 Agilent's [Mass Hunter Explorer](#) v2.0 integrates SIRIUS capabilities through a proprietary
150 client implementation. The RforMassSpectrometry initiative is currently adapting *RSirius* as
151 the foundation for *RuSirius*. *PySirius* and *RSirius* show more than fifty thousand combined
152 downloads on [Anaconda.org](#).

153 **Updates and maintainability**

154 A CI/CD pipeline automatically generates and tests the R and Python client libraries against
155 the latest SIRIUS API for each new release. The new client versions are subsequently published
156 to conda-forge. This automated workflow ensures the client libraries remain maintainable,
157 up-to-date, and resilient as new clients are added. The SIRIUS Java SDK is part of the SIRIUS
158 main application and is therefore built and distributed alongside it.

159 **Expandability and contribution**

160 The OpenAPI specification enables generation of client libraries for any [supported programming](#)
161 [language](#) of the OpenAPI Generator. We welcome community contributions of additional client
162 libraries and relegate interested readers to our [contribution guidelines](#).

163 **Installation via conda-forge**

164 Installation of either *PySirius* or *RSirius* via conda-forge automatically installs SIRIUS as a
165 dependency.

```
166 conda install -c conda-forge py-sirius-ms # Python client library (PySirius)  
167 conda install -c conda-forge r-sirius-ms # R client library (RSirius)  
168 conda install -c conda-forge sirius-ms # SIRIUS (included in the above)
```

169 **Examples and documentation**

170 Minimalistic examples illustrating the fundamental usage of *RSirius* and *PySirius* are provided
171 within the client library source code repository. Exemplary implementations demonstrating
172 the usage of individual functions corresponding to each [API module](#) are available in the test
173 suites for *RSirius* and *PySirius*. Comprehensive documentation for *RSirius* and *PySirius* is also
174 provided in the repository.

175 **Availability**

176 Both the [client library source code](#) and the [SIRIUS source code](#) are openly available on GitHub.
177 The client libraries are distributed via conda-forge as [py-sirius-ms](#) (*PySirius*) and [r-sirius-ms](#)
178 (*RSirius*). SIRIUS itself is available via conda-forge as [sirius-ms](#) or as direct [releases](#) on GitHub.
179 The Java client is hosted on a GitLab [package registry](#), and the SIRIUS API implementation is
180 part of the [SIRIUS repository](#).

181 **AI usage disclosure**

182 AI assistance was utilized for text editing of this manuscript and the [contribution guidelines](#) of
183 the repository. All changes proposed by AI systems were manually reviewed.

184 **Competing interests**

185 MF, ML, MAH, KD and SB are co-founders of Bright Giant GmbH.

¹⁸⁶ Acknowledgements

¹⁸⁷ NAH, SB, MAH and MEL are supported by the Thüringer Ministerium für Wirtschaft,
¹⁸⁸ Wissenschaft und Digitale Gesellschaft (TMWWDG) with funds from the European Union as
¹⁸⁹ part of the European Regional Development Fund (ERDF, 2023 VFE 0003 and 0029).

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