

# <sup>1</sup> scpviz: A Python bioinformatics toolkit for Single-cell Proteomics and multi-omics analysis

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## Software

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## <sup>8</sup> Summary

<sup>9</sup> Proteomics seeks to characterize protein dynamics by measuring both protein abundance and <sup>10</sup> post-translational modifications (PTMs), such as phosphorylation, acetylation, and ubiquitination, which regulate protein activity, localization, and interactions. In bottom-up proteomics <sup>11</sup> workflows, proteins are enzymatically digested into peptides that are measured as spectra, <sup>12</sup> from which these peptide-spectrum matches (PSMs) are aggregated to infer protein-level <sup>13</sup> identifications and quantitative abundance estimates. Analyzing the two levels of data at <sup>14</sup> both the peptide level (short fragments observed directly) and the protein level (assembled <sup>15</sup> from peptide evidence) in tandem is crucial for translating raw measurements into biologically <sup>16</sup> interpretable results.

<sup>18</sup> Single-cell proteomics extends these approaches to resolve protein expression at the level of individual cells or microdissected tissue regions. Such data are typically sparse, with many <sup>19</sup> missing values, and are generated within complex experimental designs involving multiple <sup>20</sup> classes of samples (e.g., cell type, treatment, condition). These properties distinguish single-cell <sup>21</sup> proteomics from bulk experiments and create unique challenges in data processing, normalization, <sup>22</sup> and interpretation. The single-cell transcriptomics community has established a mature <sup>23</sup> ecosystem for managing similar challenges, exemplified by the scanpy package ([Wolf et al., 2018](#)) <sup>24</sup> and the broader scverse ecosystem (?). Building on these foundations, scpviz extends <sup>25</sup> the AnnData data structure to the domain of proteomics. scpviz is a Python package that <sup>26</sup> streamlines single-cell and spatial proteomics workflows, supporting a complete analysis pipeline <sup>27</sup> from raw peptide-level data to protein-level summaries and downstream interpretation through <sup>28</sup> differential expression, enrichment analysis, and network analysis. The core of scpviz is the <sup>29</sup> pAnnData class, an AnnData-affiliated data structure specialized for proteomics. Together, these <sup>30</sup> components make scpviz a comprehensive and extensible framework for single-cell proteomics. <sup>31</sup> By combining flexible data structures, reproducible workflows, and seamless integration with <sup>32</sup> the AnnData, scanpy and extended scverse ecosystem, the package enables researchers to <sup>33</sup> efficiently connect peptide-level evidence to protein-level interpretation, thereby accelerating <sup>34</sup> methodological development and biological discovery in proteomics.

## <sup>36</sup> Statement of need

<sup>37</sup> Although general-purpose data analysis frameworks such as scanpy ([Wolf et al., 2018](#)) and <sup>38</sup> the broader scverse ecosystem have become indispensable for single-cell transcriptomics, <sup>39</sup> comparable tools for proteomics remain limited. Existing proteomics software often focus on <sup>40</sup> specialized tasks (e.g., peptide identification or spectrum assignment) and do not provide a <sup>41</sup> unified framework for downstream analysis of peptide- and protein-level data within single-cell

42 and spatial contexts.

43 scpviz addresses these gaps by offering an integrated system for the complete proteomics work-  
44 flow, from raw peptide-level evidence to protein-level summaries and biological interpretation.  
45 It is designed for computational biologists and proteomics researchers working with low-input  
46 or single-cell datasets from data sources such as Proteome Discoverer or DIA-NN([Demichev et al., 2020](#)).  
47

48 At the core of scpviz is the pAnnData class, an AnnData-affiliated data structure specialized  
49 for proteomics. It organizes peptide (.pep) and protein (.prot) AnnData objects alongside  
50 supporting attributes such as .summary, .metadata, .rs matrices (protein-peptide relation-  
51 ships), and .stats. This design allows users to move flexibly between peptides and  
52 proteins while maintaining compatibility with established Python libraries for data science and  
53 visualization.

54 Beyond data organization, scpviz implements proteomics-specific operations, including filter-  
55 ing (e.g., requiring proteins supported by at least two unique peptides), normalization and  
56 imputation methods tailored for sparse datasets, and visualization tools such as PCA (Principal  
57 Component Analysis), UMAP (Uniform Manifold Approximation and Projection for Dimension  
58 Reduction), clustermaps, and abundance plots. For downstream interpretation, it integrates  
59 with UniProt for annotation and string-db for enrichment and network analysis (?; ?; ?). The  
60 framework also incorporates single-cell proteomics-specific normalization strategies such as  
61 directLfq ([Ammar et al., 2023](#)), ensuring robust quantification across heterogeneous samples.  
62 Finally, pAnnData objects interface seamlessly with scanpy ([Wolf et al., 2018](#)) and other  
63 ecosystem tools such as harmony ([Korsunsky et al., 2019](#)), enabling direct incorporation into  
64 established single-cell workflows.

65 The design philosophy of scpviz emphasizes both usability and extensibility. General users can  
66 rely on its streamlined API to import, process, and visualize single-cell proteomics data without  
67 deep programming expertise, while advanced users can extend the framework to accommodate  
68 custom analysis pipelines. The package has already been applied in published papers and  
69 preprints ([Dutta et al., 2025](#); [Pang et al., 2025](#); [Uslan et al., 2025](#)) as well as manuscripts  
70 in preparation, and it has been incorporated into graduate-level training to illustrate how  
71 proteomics workflows parallel to those in single-cell transcriptomics.

72 The applications of scpviz span diverse areas of life sciences research, from studying protein  
73 dynamics and signaling pathways to integrating proteomics with transcriptomics for multi-  
74 omics analysis. By bridging the gap between raw mass spectrometry data and systems-level  
75 interpretation, scpviz provides a versatile and reproducible platform for advancing single-cell  
76 and spatial proteomics.

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