

# Mass Spectrometry-based Multi-Omics: Combined Studies of Proteomics, Metabolomics, and/or Lipidomics

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## Abstract

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Studies that integrate unbiased measurements across at least two omics layers are often referred to as “multiomics”. Measurable “omes” include the genome, transcriptome, proteome, and metabolome. Any combination of omic measures can be referred to as multiomics; for the scope of this review we focus on research combining proteomics and metabolomics. Mass spectrometry is the leading technique for analysis of the proteome and the metabolome. Due to improvements in sample preparation and data collection, more studies are incorporating both mass spectrometry-based proteomics and metabolomics. In this review, we discuss the perceived value of multiomics, advances in sample preparation and data collection, the current state of multiomic data integration, and clinical examples of multiomic analysis. Finally, we explore major barriers preventing democratization of mass spectrometry based multiomics to the same level as nucleic acid analysis, and we suggest solutions to break these barriers.

# Introduction

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A major goal of biomedical research is understanding how changes in biomolecule compositions of cells and tissues lead to disease phenotypes. The genome serves as a library of possible transcripts, some of which are instructions for proteins, and proteins act on metabolites. The genome thus indirectly determines the set of possible cellular phenotypic states, and the exact cellular state at any given time depends on interaction between the genome with endogenous and exogenous environmental cues. To understand how the genome connects to phenotype, measurement of the genome must be accompanied by measurement of downstream layers of the central dogma of biochemistry (Figure 1?).

Most RNAs encode for proteins, and aside from noncoding RNA and ribozymes, most of the potential functions encoded in a genome are carried out by proteins. Although extensive research has nearly commoditized RNA measurement or transcriptomics [1], multiple studies have found poor correlation between proteins and their corresponding mRNA [2]. Therefore, functional understanding of a biological state demands quantitations of all proteins that are present.

Large scale study of the proteins in a biological system is known as proteomics [3]. Proteomics originally meant large scale studies of protein quantities [4], but it has grown to include large scale studies of protein structure[5] [6] [7], protein-small molecule interactions[8], and protein-protein interactions [9]. Proteomics is increasingly applied to clinical studies to gain understanding of disease mechanisms in humans [10].

Many argue that quantities of proteins are not sufficient to enable understanding and dissection of biological systems. The presence of a protein does not necessarily equal the presence of that protein's function. Protein functions are regulated in many ways, including allosteric feedback by metabolites and post-translational modifications such as phosphorylation, acylation, or proteolysis [11]. Roughly 2,700 cellular proteins are predicted to be enzymes [12], many of which act on metabolites that are intermediates required for cell proliferation or energy production.

The large scale measurement of the cellular metabolite pools is known as metabolomics [13], which is sometimes further separated into polar metabolomics and lipidomics [14]. Due to the chemical heterogeneity of metabolites, metabolomics studies use a variety of different separation and mass spectrometry (MS) techniques, including both gas chromatography (GC) [15] and liquid chromatography (LC)[16]. The field of metabolomics is trending toward more use of LC-MS over NMR [17].

Measurement of multiple omic layers is known as multiomics. MS measures peptides (or intact proteins [18]) for proteomic analysis, but MS also measures metabolites for metabolomic analysis. Increasingly studies use MS to measure both proteomes and metabolomes from the same sample. Although multiomics can refer to measuring at least any two omes, throughout this review we focus on multiomic measurements between proteomes and metabolomes.

Integrating data from multiomics should provide more information about the cellular state than the sum of each dataset. Multiomic data integration methods have been developed with different classes of goals. More work is needed to develop methods that take advantage of multiomic data to discover new biological insights into how systems work.

Previous reviews have covered related topics. A recent review surveyed multiomic studies in mitochondria [19] with a focus on sample preparation, quality control, methods to collect MS data, and linking measured molecules to phenotypes of respirometry. Another review gives an overview of

each omic method, considerations before performing a multiomic method, and multiomic integration [20]. Misra *et al.* reviewed multiomics with a focus on multiomic integration tools [21]. A metabolomics-centered review was recently published that discusses multiomic integration [22].

In this review we focus on the following topics with a focus on MS-based multiomics studies that combine proteomics and metabolomics:

- sample preparation methods
- data collection methods
- data integration
- applications to model organisms
- applications to clinical studies

List of Planned Figures:

1. overview of how omic layers are related showing different ‘flavors’ of each omic analysis

genomics: transcription factor binding (chip-seq), long range structure (ATAC seq?), etc  
transcriptomics: microarrays, RNA-seq, long read nanopore or pacbio  
proteomics: interactomics, structural proteomics (thermal proteome profiling or CETSA), PTMomics  
metabolomics: polar metabolomics, lipidomics,

2. Workflow for typical MS-based multi-omics. from sample preparation, extraction to data collection, data integration and processing.
3. Table of clinical multiomics studies
4. Figure showing concepts in multiomics integration?
- 5.

## Sample Preparation for Multi-Omic Analysis

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The basics: As with any methodology, the very first but also a crucial step for analytical success is proper sample preparation. This is particularly vital for mass spectrometry-based multi-omic analysis, the selected sample preparation strategy is a key determinant for information that will be obtained. Given the great diversity and complexity of biological samples being tested in proteomics and metabolomics, even subtle differences in sample preparation methods can have profound effects on the types of molecules being extracted, which further affect the retention time, signal stability and ionization efficiency. For example, analysis of different types of biomolecules requires specific sample extraction procedures. The most efficient extraction protocols often only cover a restricted type of biomolecules due to their different physicochemical properties. After any necessary dissection or collection, biological tissue samples require a step to

### 1, Sample preparation for proteomics

Protein preparation for MS analysis can be accomplished by many methods, so it is important to understand the steps leading to analysis. While intact proteins are typically studied by gel electrophoresis, the most common mass spectrometry workflows for complex protein samples analyze peptides, which are easier than proteins to fractionate by LC. Peptides also ionize and fragment more efficiently than whole proteins, resulting in spectra that are easier to interpret for protein identification. Peptide preparation involves reduction and alkylation of cysteines, digestion of

the sample into peptides, desalting and concentration of the peptides and final analysis of these peptides by ionization (e.g., ESI) plus orbitrap-based MS.

## 2, Sample preparation for metabolomics

### 2.1 non-targeted metabolomics

[\[23\]](#)

### 2.2 targeted metabolomics

### 2.3 lipidomics

[\[24\]](#)

## 3, Integrative sample preparation for multi-omics

In the context of multi-omics analyses, being able to perform multiple measurements on the same sample can also decrease experimental variation. In this section, we will review several integrative sample preparation technical advances that increased the capabilities of multiomic analysis. One important goal is to decrease the variation between different samples, ultimately to .

[\[25\]](#)

[\[26\]](#)

-“An Integrated, High-Throughput Strategy for Multiomic Systems Level Analysis” [\[27\]](#)

-“Detergent-Free Simultaneous Sample Preparation Method for Proteomics and Metabolomics” [\[28\]](#)

-“MPLEx: a Robust and Universal Protocol for Single-Sample Integrative Proteomic, Metabolomic, and Lipidomic Analyses” [\[29\]](#)

-“Single-platform ‘multi-omic’ profiling: unified mass spectrometry and computational workflows for integrative proteomics–metabolomics analysis” [\[30\]](#)

-“Three-in-One Simultaneous Extraction of Proteins, Metabolites and Lipids for Multi-Omics” [\[31\]](#)

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## New developments of mass spectrometry-based methods for multi-omics

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### 4.1 proteomics

#### 4.1.1 Traditional standard methods for proteomics

(Remember to mention here)

#### 4.1.2 Direct infusion methods for proteome analysis (high-throughput methods)

For current proteomic analysis methods, time-consuming chromatographic separation (typically requiring 30–60 min per sample or even longer) is required to protect the coverage, repeatability, robustness and quantification ability. However, with the rapid application of multiomics results in drug development, biomarker discovery studies and clinical diagnosis. High-throughput methods is highly desirable to boost these fields forward.[32]

(as a high-throughput method, MALDI based proteome analysis should be mentioned here, for example, the application of MALDI for identification of species of bacteria and fungus through their specific peptides)

#### 4.2 metabolomics

To accurately and reliably interpret data derived from metabolomics and lipidomics studies, enormous mass spectrometry based methods were developed during the past decades. (remember to mention the application of MALDI for metabolites analysis, although the drawback of MALDI-tof is obvious.)

Drawbacks: 1,the background of organic matrix in the low molecular weight region 2,the obtained information of MALDI is still very limited, no more than 300 identified metabolites, and also quantification is difficult.

3,as a non-consistent ion source, currently TOF is the typical mass analyser for MALDI, which still suffers from relative low resolution. FTICR can connect MALDI )

##### 4.2.2 Direct infusion mass spectrometry methods for high-throughput analysis of metabolites.()

(direct infusion and so called flow injection MS. do not know the differences, seems saying the same thing.) In typical metabolomic analysis workflow, MS is typically coupled with liquid or gas chromatography, enables separating metabolites within a complex sample and further increases the overall sensitivity by minimizing the ion suppression effect. However, this approach also bring for metabolomics analysis is that chromatographic separation is time-consuming (20–60 min per sample), preventing its application for high-throughput metabolomic screens of large scale samples that required for biomarker discovery studies and clinical analysis.

[24]

[24]

#### 4.4 Integrated Strategy for Mass Spectrometry-Based Multiomics

Mass spectrometry (MS) serves as the centerpiece technology for proteome and metabolome analysis. To gain a better understanding of the multifaceted networks of myriad actions in complex organisms, integration of different multiomic layers is increasingly explored such as joint methods of different omics.

“Multiomics Method Enabled by Sequential Metabolomics and Proteomics for Human Pluripotent Stem-Cell-Derived Cardiomyocytes” [doi?]

[33]

[34, =pdf]

[35, =pdf]

## Multi-Omic Data Integration

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Using mass spectrometry, the data collected is proteomics and metabolomics. Integration of these datasets

Reviews:

Multi-omics Data Integration, Interpretation, and Its Application [36]

Network approaches to systems biology analysis of complex disease: integrative methods for multi-omics data [37]

Argonaut data integration manuscript - [38]

Multi-omics data integration considerations and study design for biological systems and disease - [URL? <https://doi.org/10.1039/D0MO00041H>]

List of software available for multiomic (proteomic/metabolomic) integration

Metaboanalyst 4.0 - Metabolomics data analysis, interpretation, and integration with other omics data [@PMID: 31756036]

Paintomics 3.0 (web based) - Joint visualization of transcriptomics and metabolomics data [@URL: <https://doi.org/10.1093/bioinformatics/btq594>]

integrOmics (R package) - Integrative analysis of two types of omics datasets [@URL: <https://doi.org/10.1093/bioinformatics/btp515>]

Omics Integrator - Maps protein data to other data sets [@URL: <https://doi.org/10.1371/journal.pcbi.1004879>]

mixOmics (R package) - Data exploration, dimension reduction, and visualization [@URL: <https://doi.org/10.1371/journal.pcbi.1005752>]

"Integration strategies of multi-omics data for machine learning analysis" [39]

Split strategies into five categories

Early integration: Place all omics datasets into single matrix

tends to focus on largest dataset, missing data from other omic datasets  
[@PMID:18178960]

simple, easy implementation, allows ML models to learn interactions between layers

Mixed integration: Map each dataset into new representation before combining

Intermediate integration: Transform datasets into common and omic-specific representations

Late Integration: analyze each omic layer separately and combine final predictions

hierachical integration: base integration of datasets on regulatory relationships between omic layers

"Multi-omics data integration considerations and study design for biological systems and disease" [40]

mixomics R tools [41/journal.pcbi.1005752]

"Multi-omics Data Integration, Interpretation, and Its Application" [42]

Correlation analysis: Network Analyses and Data Integration of Proteomics and Metabolomics From Leaves of Two Contrasting Varieties of Sugarcane in Response to Drought [URL? https://internal-journal.frontiersin.org/articles/10.3389/fpls.2019.01524/full] Regularized Canonical Correlation Analysis

Clustering and variable selection evaluation of 13 unsupervised methods for multi-omics data integration [43, =true]

Data integration and predictive modeling methods for multi-omics datasets [URL? https://pubs.rsc.org/en/content/articlehtml/2018/mo/c7mo00051k]

Specifically proteomics and metabolomics: [44] Uses feature selection for proteomic and metabolomic data to create logistic regression predicting survivors from non-survivors during severe septic shock

In Cambiaghi et al 2018 [44], targeted metabolomics and untarged proteomics were collected from septic shock patients from day one to day seven. To integrate the data, logistic regression models were built to link the metabolomics and proteomics data with patient outcome. Feature selection was performed on the metabolomic data set with minimal-redundancy-maximal-relevance (mRMR) [PMID:16119262]. Further feature selection was performed using an elastic net logistic model to select variables up to the sample size of 17. Linear Discriminant analysis and partial least squares discriminant analysis were used on the top scoring metabolites. To integrate targeted metabolomics with untargeted proteomics, mRMR was applied. The top 50 ranked proteins were combined with top 50 ranked metabolites and mRMR repeated. Additionally, clinical data was integrated in a similar manner. This method represents an early integration strategy with feature selection as the means to determine biological relationships.



[45/] Integration of Metabolomic and Proteomic Phenotypes - correlation and protein metabolite covariance In Wienkoop et al 2008

[46] Integration of Proteomics and Metabolomics Revealed Metabolite-Protein Networks in ACTH-Secreting Pituitary Adenoma - hierarchical clustering, pathway enrichment analysis In Feng et al 2018, metabolomics data and proteomics data were collected on adrenocorticotrophic hormone-secreting pituitary adenomas. Metabolomics data was prepared and analyzed with MetaboAnalyst [47]. Hierarchical cluster analysis was performed with MetaboAnalyst. Proteomics data was analyzed with ProteinPilot and searched against the SwissProt Database. Protein enrichment analysis was performed with Gene Set Enrichment Analysis [48]. Further integration such as network analysis was performed in MetaboAnalyst. This represents a late integration strategy where each dataset is processed separately and conclusions drawn from the connections linking the separate analyses.

[49] - Integrated Datasets of Proteomic and Metabolomic Biomarkers to Predict Its Impacts on Comorbidities of Type 2 Diabetes Mellitus - "ingenuity pathway analysis" In Cheema et al 2020,

[50] - Integrated Proteomics and Metabolomics Analysis Highlights Correlative Metabolite-Protein Networks in Soybean Seeds Subjected to Warm-Water Soaking - hierarchical clustering of differentially expressed proteins - "MapMan" - GO enrichment In Min et al 2020, proteomics and metabolomics data was collected on soybean seeds soaked at 4, 25, and 55C. Proteomics data was processed using MaxQuant and Perseus. Enrichment of significantly changed proteins was analyzed using MapMan[cite]. Metabolomic data was analysed by unit-variance scaling the data and using principal component analysis and partial least-squares discriminant analysis using Soft Independent Modeling of Class Analogy (SIMCA)[cite]. MetaboAnalyst was used for Hierarchical Cluster analysis between samples and compounds. Pearson's correlation analysis was performed with Statistical Analysis System (SAS)[cite]. This represents a late integration strategy where each dataset is processed separately and conclusions drawn from the connections linking the separate analyses.

How are people analyzing data? What software is available for combinational analysis? Why aren't these being utilized? What gaps in analysis are there in the software?

Metabolomics and Proteomics represent two facets of the spectrum of multiomic data analysis. Through the integration of these datasets, insights can be gleaned that may not be evident from each individual source. The strategies for the integration of any set of multiomic data can be broadly divided into two categories, namely early integration and late integration[source]. Early integration strategies combine the datasets, potentially with adjustments to make the datasets compatibly scaled, before further analysis [Source]. The essence of late integration strategies is to analyze each dataset separately then combine knowledge gained from each to support further conclusions [Source]. Each strategy employed have positive and negative aspects that affect the downstream analysis of the multiomic data. The

## Mass Spectrometry-based Multi-Omics Applied to Model Organisms

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"Multiomics Method Enabled by Sequential Metabolomics and Proteomics for Human Pluripotent Stem-Cell-Derived Cardiomyocytes" [51]:

Multi-omics Reveal Specific Targets of the RNA-Binding Protein Puf3p and Its Orchestration of Mitochondrial Biogenesis [52]

Mitochondrial protein functions elucidated by multiomic mass spectrometry [53]

Multi-omic mitoprotease profiling reveals role for oct1p [\[54\]](#)

“An integrative systems genetic analysis of mammalian lipid metabolism” [\[55\]](#) \* proteomics and lipidomics of mouse liver across 107 genetically different strains \*

## Clinical applications of multi-omics (proteomics, metabolomics, and/or lipidomics)

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Literature currently focused on proteogenomics (integration of genome, transcriptome, and proteome) “Clinical multi-omics strategies for the effective cancer management” [\[56\]](#)

Holistic approach has the ability to improve prognostics and predictive accuracy of disease phenotypes to aid in better treatment and prevention “Multi-omics Data Integration, Interpretation, and Its Application” [\[57\]](#)

## Approaches in cancer research:

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- “Integration of Proteomics and Metabolomics Revealed Metabolite–Protein Networks in ACTH-Secreting Pituitary Adenoma” [\[56\]](#): gas chromatography-mass spectrometry (GC-MS) for metabolomics, plus nano liquid chromatography tandem-mass spectrometry (nanoLC-MS/MS) for proteomics. Using metabolomic and proteomic data to identify signaling pathways important in metabolic regulation of tumorigenesis. Reveal biomarkers for disease diagnosis, monitoring and therapeutic targets. Looking for changes in tumor on the pituitary gland (adrenocorticotrophic hormone-secreting pituitary adenomas (ACTH-PA)). Downregulated glycolysis and fatty acid synthesis. Myc signaling pathway significantly participated in the metabolic changes and tumorigenesis of ACTH-PA.
- “L-Arginine Modulates T Cell Metabolism and Enhances Survival and Anti-tumor Activity” [\[58\]](#): investigation of proteomic and metabolic adaptations underlying CD4<sup>+</sup> T cell activation using MS. Proteins were analyzed by liquid chromatography-coupled mass spectrometry (LC-MS). In parallel, polar metabolites were analyzed by non-targeted flow-injection metabolomics (semi-quantitative method that allows rapid and deep profiling of metabolites. limitations: isobaric compounds cannot be discriminated + possible in-source degradation). Findings: increased L-arginine levels has effects on T cell activation, differentiation, and function.
- “Concurrent lipidomics and proteomics on malignant plasma cells from multiple myeloma patients: Probing the lipid metabolome” [\[59\]](#) comparing targeted and untargeted lipidomics and proteomics between relapsed and newly diagnosed MM patients
- “Proteometabolomics of bladder cancer: Current and future prospects” [\[doi:10.3233/CBM-150479?\]](#): simultaneous proteome and metabolome data from urine and blood for urinary bladder cancer patient surveillance to aid in early detection of bladder cancer
- “Grade-Dependent Metabolic Reprogramming in Kidney Cancer Revealed by Combined Proteomics and Metabolomics Analysis” [\[60\]](#)

## Approaches in COVID-19 research:

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- “Proteomic and Metabolomic Characterization of COVID-19 Patient Sera” [\[61\]](#): proteomic and metabolomic profiling of serum from 46 COVID-19 patients. Stable isotope labeled proteomics TMTpro (16plex) and ultra performance liquid chromatography/tandem mass spectrometry (UPLC-MS/MS) untargeted metabolomics to analyze samples. Dysregulation of macrophage, platelet degranulation, complement system, and metabolic suppression in severe COVID-19 patients, useful for selection of biomarkers for severity evaluation. Implemented machine learning model using expression levels of 22 serum proteins and 7 metabolites with overall accuracy of

93.5% in the training set. \*Large-Scale Multi-omic Analysis of COVID-19 Severity: [62]: investigation of proteins, lipids, and metabolites associated with COVID-19 status and severity in pts admitted with moderate to severe respiratory issues presumably related to SARS-CoV-2 infection.

- “Metabolomic/lipidomic profiling of COVID-19 and individual response to tocilizumab” [63]
- “Multi-Omics Resolves a Sharp Disease-State Shift between Mild and Moderate COVID-19” [64]
- “Multi-Omics integration analysis of respiratory specimen characterizes baseline molecular determinants associated with COVID-19 diagnosis” [65]
- “Longitudinal multi-omics analysis identifies responses of megakaryocytes, erythroid cells and plasmablasts as hallmarks of severe COVID-19 trajectories” [66]

## Misc disease states:

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- “Integration of metabolomics and proteomics in multiple sclerosis: From biomarkers discovery to personalized medicine” [67]: majority of biomarkers for MS are not sensitive and/or specific enough to be used for population screening. From the 188 proposed biomarker candidates for MS found in CSF, only 10 (5%) have been successfully verified, while 20 have been falsified. This is likely due to the inter-individual variation. Integrating omics approaches may help with this pitfall; proteomics in combination with the other “omics” may synergize to provide a more precise spectrum of information capable of better characterizing a phenotype. When proteins alone, metabolites alone and proteins and metabolites together were uploaded into the ingenuity pathway analysis (IPA) tool, number of matched molecules for “proteins and metabolites” is higher than the sum of molecules matched in the separate searches.
- “High-throughput mediation analysis of human proteome and metabolome identifies mediators of post-bariatric surgical diabetes control” [68] patients with type 2 diabetes randomized to RYGB vs nonsurgical diabetes/weight management; fasting plasma proteome and metabolome were assayed up to 3 years. Plasma proteome profiling was performed using the high-throughput DNA aptamer-based SOMAscan assay platform. Plasma metabolomics were profiled using a commercial semi-quantitative mass spectrometry-based platform (Metabolon, Inc.) using UPLC-MS/MS. Main mediator of improved glycemic control was GHR, which was reduced 3 months after RYGB. GH signaling = diabetogenic effects.

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