

# Comparative Analysis of High-Throughput Metaproteomics Data in Unipept

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# Acknowledgements



# Summary





# Samenvatting



# List of publications



# List of conferences



# **Chapter 1**

## **Introduction**





## **Part I**

# **Unipept Desktop**



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The introductory text for this part comes here...

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## Chapter 2

# Unipept Desktop: a faster, more powerful metaproteomics analysis tool

**Abstract** Metaproteomics has become an important research tool to study microbial systems, which has resulted in increased metaproteomics data generation. However, efficient tools for processing the acquired data have lagged behind. One widely used tool for metaproteomics data interpretation is Unipept, a web-based tool that provides, amongst others, interactive and insightful visualizations. Due to its web-based implementation, however, the Unipept web application is limited in the amount of data that can be analyzed. In this manuscript we therefore present Unipept Desktop, a desktop application version of Unipept that is designed to drastically increase the throughput and capacity of metaproteomics data analysis. Moreover, it provides a novel comparative analysis pipeline and improves the organization of experimental data into projects, thus addressing the growing need for more performant and versatile analysis tools for metaproteomics data.

## 2.1 Introduction

Metaproteomics is a relatively young research field that focuses on the study of microbial environments and complex ecosystems, and of the interactions between the organisms involved, through the analysis of the proteins extracted from these environments. Over the past years, the technology to identify proteins from such complex samples has been greatly improved, allowing metaproteomics to transition from relatively small studies to large scale experiments (Rechenberger *et al.*, 2019; Wilmes *et al.*, 2015). The key enabling technologies for this transition are improved mass spectrometers and more powerful proteomics approaches, which have both come a long way since the introduction of metaproteomics analysis in 2004 (Rodríguez-Valera, 2004; Yates, 2019). To allow efficient processing of the resulting increase of acquired data, various dedicated tools have been made available to support metaproteomics data analysis (Muth *et al.*, 2015; Van Den Bossche *et al.*, 2020), but even with this increased bioinformatics support, many challenges still need to be overcome, especially regarding downstream analysis of the obtained identifications (Schiebenhoefer *et al.*, 2019).

Unipept is a leading tool for such downstream metaproteomics data analysis (Herbst *et al.*, 2016) that currently consists of a web application (Gurdeep Singh *et al.*, 2019), a web service, and a command line tool (Verschaffelt *et al.*, 2020). The Unipept web application provides users with the ability to analyze a metaproteomics sample and extract taxonomic and functional information from environmental samples derived from a variety of origins, ranging from the human gut to biogas plants. The Unipept web application provides users with interactive visualizations and allows them to, for example, filter out all functions that are associated with a specific taxon. Due to its web-based nature, however, the size and number of samples that can be analyzed by Unipept are limited. And while it is currently possible to analyze larger data sets using the Unipept CLI, this requires more sophisticated bioinformatics skills and does not provide the interactive link between taxa and functional annotations.

Because of the browser limitations, it can already take a substantial amount of time to process relatively small samples (e.g. containing up to a few thousand identified peptides) using Unipept, depending on the specific search configuration used. These limitations have become an issue, as the advances in metaproteomics have not only increased

data set sizes, but have also increased the number of data sets that need to be processed (Zhang and Figeys, 2019).

In order to accommodate this evolution, the throughput of metaproteomics data analyses needs to increase as well, in turn requiring tools that are not constrained in the amount of memory and CPU resources they are allowed to consume. Moreover, analysis results also need to be retained for future reference, ideally in a project-based approach that can group multiple samples, and the corresponding results should be easily shareable with other researchers.

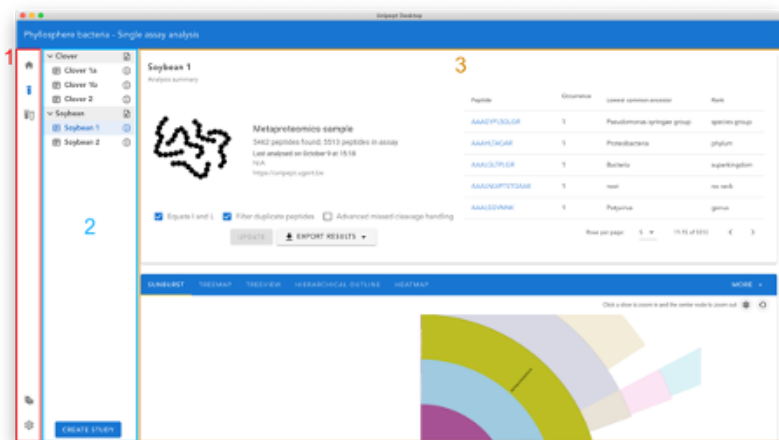
For specific applications, it is also important that all data is processed offline or on-site rather than being sent over the internet. For instance, sensitive medical data is often not allowed to be sent to external services for processing, but must be kept in-house to safeguard patient confidentiality and privacy.

All of the above issues need to be resolved in order to support the growing interest in, and reach of, metaproteomics. We therefore here present the Unipept Desktop Application, a novel cross-platform desktop application designed to specifically overcome these challenges while also retaining the functionality that exists in the current web app.

## 2.2 Implementation

The Unipept desktop application provides three different types of analyses: *i*) single assay analysis, *ii*) inter-assay comparative analysis, and *iii*) tryptic peptide analysis. The single assay analysis performs a full taxonomic and functional analysis of a single assay and corresponds to the default “metaproteomics analysis” as presented by the Unipept web application. The inter-assay comparative analysis on the other hand, provides the ability to explore similarities and differences between multiple assays. While the comparison of multiple assays was already possible with the Unipept web application, this was only available for a limited number of quite small assays due to strict memory constraints posed by web browsers. The tryptic peptide analysis, lastly, can be used to look up which proteins, taxa and functions are associated with a given peptide.

Unipept Desktop delivers these core functions through a concise user interface (Figure 2.1) that consists of three main parts: the sidebar,



**Figure 2.1:** Screenshot of the Unipept Desktop application. The analysis page of the desktop application is depicted here and consists of three main parts: the sidebar that is used to navigate between the different analysis pipelines and functions of the application (1), the project explorer that displays a hierarchical view of the project (2) and the content view that renders analysis results (3).



the project explorer, and the content view. The sidebar on the far left allows the user to navigate between the different analysis pipelines and functions of this application. Directly to the right of the sidebar is the project explorer that allows the user to switch between assays, and to modify the project. The project explorer is only shown when performing single assay or comparative analyses. Assays and studies can be renamed or deleted by right clicking them, after which a context menu opens. Lastly, the content view takes up most of the application's visual space and presents either analysis results or the settings page.

The Unipept Desktop Application also allows offline analysis of data through a choice of the API endpoint in the settings menu. This endpoint, which uses the Unipept API and by default connects to the online Unipept system, can be configured to call any service that supports the Unipept API. By setting up a local instance of the Unipept backend system, the user can thus ensure that all data remains locally. Setting up a local Unipept back-end is possible by cloning the open source Unipept repository on GitHub, but requires advanced technical knowledge. We plan to make the installation process of these custom API endpoints even easier with future releases of Unipept.

Unipept Desktop is powered by the cross-platform Electron framework, which in itself is powered by Chromium browser technology. This means that the application is developed with web-centric technologies, such as the Vue frontend framework and TypeScript, and hence we were able to reuse large parts of the web app's codebase. The choice for the Electron platform was mostly driven by the extensive suite of different functionalities that can be integrated with minimum configuration efforts. Thanks to the Electron platform we can provide an automatic update mechanism, easily generate installation packages for all major platforms (Windows, macOS and Linux), and include automatic crash reporting, amongst others. Once installed, the Unipept Desktop application can thus update fully autonomously in the background, ensuring that users always have the latest functionality and bug fixes installed.

### **2.2.1 Project-centric analysis**

The Unipept Desktop Application has full access to the local filesystem. Hence, it can store an arbitrary amount of data and does not need to worry about strict size limits; this in contrast to web applications that are only allowed to store up to a few megabytes using the local

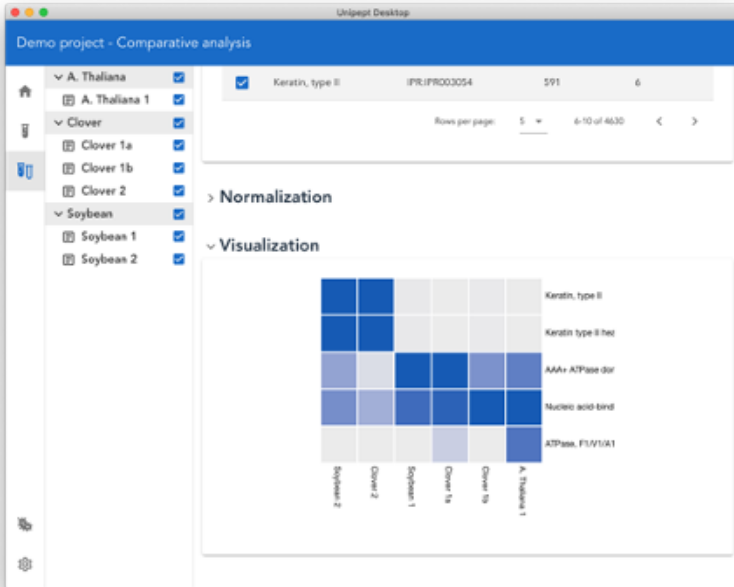
storage API. This allows us to improve upon the organization of data sets by introducing project-based data management capabilities. In accordance with the terminology introduced by the ISA-tab standard for experimental metadata annotation (Sansone *et al.*, 2012), we now refer to a data set derived from a sample as an “assay”, while a study is a grouping of multiple, related assays, and a Unipept project represents a collection of such studies.

On the file system, a project is stored in a single folder that contains an SQLite database file, a subfolder for each study and one text file per assay, located in the subfolder of the corresponding study. This folder can be modified outside of the application, using the default file explorer application of your operating system, thus providing maximum flexibility. All changes made to this project folder are automatically detected and imported by the application, granting users the ability to mass import assays and edit project properties with external applications. The application accepts simple text files with one peptide per line. In order to quantify peptide occurrence, a peptide can be included more than once in this file and the “filter duplicate peptides” option should be disabled for the analysis.

Because projects are folder-based, they can contain both the raw input data as well as the analysis results for an assay, making it practical for users to share projects with each other, for instance, in the form of compressed project folders. In addition, previously performed analyses do not need to be recomputed when the application is restarted, as opposed to analyses that were run on the Unipept website, which need to be recomputed every time the web site is closed.

### **2.2.2 Comparative analysis**

The Unipept Desktop Application provides both intra-assay and inter-assay comparative analyses that are rendered as heatmap visualizations. The intra-assay comparison can be started from the single assay analysis page by selecting the heatmap tab and provides a wizard to guide users through the set-up process of the comparison (Figure 2). Users are required to select two types of data sources (one for each axis of the heatmap) and indicate which items should be compared. Four different data sources are currently supported: NCBI taxa, GO terms (The Gene Ontology Consortium, 2019), EC numbers and InterPro entries (Finn *et al.*, 2017).



**Figure 2.2:** Screenshot of the inter-assay comparative analysis pipeline. Note that it is possible to select multiple assays from the project explorer. A heatmap is constructed from the set of items that were selected for comparison at the top of the page.

The inter-assay comparative analysis is designed to visualize differences and similarities in functional or taxonomic composition of multiple assays. Here too, users are presented with a wizard that is similar to the one found in the intra-assay comparison. For inter-assay comparisons, however, the horizontal axis of the heatmap is reserved for the set of selected assays, and users can therefore only select one collection of items that should be compared between the different assays.

Because the number of peptides can drastically differ between multiple assays, three different normalization techniques are provided to the user. The default setting normalizes the heatmap globally, i.e. the minimum and maximum values over the complete grid are computed and all grid values are normalized with respect to these values. The other two normalization techniques also normalize based on minimum and maximum values, but restricted within a row or column, respectively.

It is worth noting that, while the comparative analysis pipeline was originally designed for the Unipept Desktop Application, a slimmed-down version has meanwhile also been integrated into the Unipept web app.

With the advent of the Unipept Desktop Application, users now have a variety of ways in which they can use Unipept. A comparison between the various functionalities offered by these different services is provided in Table 1 below.

## 2.3 Conclusion

Unipept Desktop is a novel desktop application that extends upon the Unipept web application by eradicating the strict limitations posed by the web-based nature of this application to increase metaproteomics data analysis throughput. Moreover, the Unipept Desktop Application adds new features such as allowing users to structure their data in a hierarchical project-based system, to keep track of their analysis results, and to share or distribute these results very easily. Whereas the Unipept web application is limited to assays with up to 50 000 peptides, the Unipept Desktop Application supports assays containing one million peptides or more. For reference, the desktop app can analyze between 250 and 2000 peptides per second (without advanced missed cleavage handling enabled), depending on the type of assay that's being analyzed.

In a future release of the Unipept Desktop Application, we plan to provide support for the preparation of custom reference databases and further improve support for offline analysis. This will allow us to gradually evolve to a tool that is not only suitable for metaproteomics data analysis, but also for novel proteogenomics analysis techniques for complex environmental samples.

Our choice for the Electron framework proves to be very valuable as well, as a large portion of Unipept's codebase can thus be shared between the new desktop application and the existing web application. This in turn allows us to easily migrate (a slimmed-down version of) specific desktop features to the web app, and vice versa.

## 2.4 Availability

The source code for Unipept Desktop is open source and provided under the MIT license as a repository on GitHub: <https://github.com/unipept/unipept-desktop>. Pre-generated installers for Windows, macOS and Linux (AppImage format) can be downloaded from the release page of our GitHub repository. Installation instructions and documentation for the Unipept Desktop Application can be found on our website: <https://unipept.ugent.be/desktop>.

## 2.5 Acknowledgements

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## **Chapter 3**

# **Support for novel proteogenomics analysis in Unipept**





## **Part II**

# **Other projects**



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## Chapter 4

# MegaGO: a fast yet powerful approach to assess functional Gene Ontology similarity across meta-omics data sets

**Abstract** The study of microbiomes has gained in importance over the past few years and has led to the emergence of the fields of metagenomics, metatranscriptomics, and metaproteomics. While initially focused on the study of biodiversity within these communities, the emphasis has increasingly shifted to the study of (changes in) the complete set of functions available in these communities. A key tool to study this functional complement of a microbiome is Gene Ontology (GO) term analysis. However, comparing large sets of GO terms is not an easy task due to the deeply branched nature of GO, which limits the utility of exact term matching. To solve this problem, we here present MegaGO, a user-friendly tool that relies on semantic similarity between GO terms to compute the functional similarity between multiple data sets. MegaGO is high performing: Each set can contain thousands of GO terms, and results are calculated in a matter of seconds. MegaGO

is available as a web application at <https://megago.ugent.be> and is installable via pip as a standalone command line tool and reusable software library. All code is open source under the MIT license and is available at <https://github.com/MEGA-GO/>.

## 4.1 Introduction

Microorganisms often live together in a microbial community or microbiome where they create complex functional networks. These microbiomes are therefore commonly studied to reveal both their taxonomic composition as well as their functional repertoire. This is typically achieved by analyzing their gene content using shotgun metagenomics. Whereas this approach allows a detailed investigation of the genomes that are present in such multiorganism samples, it reveals only their functional potential rather than their currently active functions (Jansson and Baker, 2016). To uncover these active functions within a given sample, the characterization of the protein content is often essential (Lohmann *et al.*, 2020).

The growing focus on functional information as a complement to taxonomic information (Louca *et al.*, 2016) is derived from the observation that two taxonomically similar microbial communities could have vastly different functional capacities, whereas taxonomically quite distinct communities could have remarkably similar functions. Whereas the investigation of the active functions is thus increasingly seen as vital to a complete understanding of a microbiome, the identification and comparison of these detected functions remains one of the biggest challenges in the field (Schiebenhoefer *et al.*, 2019).

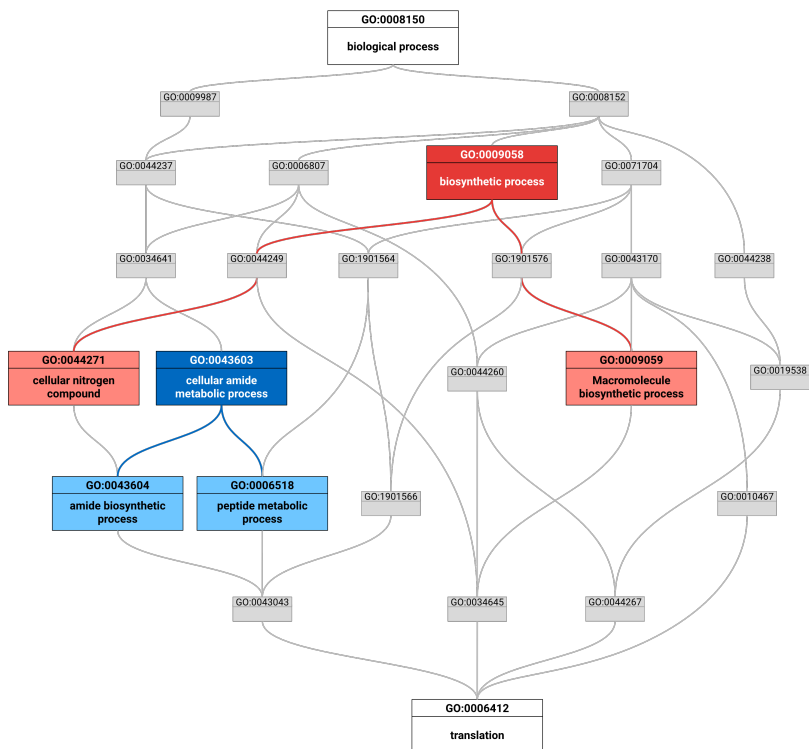
Several omics tools exist to describe functions in microbial samples, although these tools link functionality to different biological entities such as genes, transcripts, proteins, and peptides (Muth *et al.*, 2015, 2018; Van Den Bossche *et al.*, 2020; Verschaffelt *et al.*, 2020; Gurdeep Singh *et al.*, 2019; Riffle *et al.*, 2018; Schneider *et al.*, 2011; Schiebenhoefer *et al.*, 2020; Huerta-Cepas *et al.*, 2019; Huson *et al.*, 2007). However, most tools are capable of directly or indirectly reporting functional annotations as a set of Gene Ontology (The Gene Ontology Consortium, 2019) (GO) terms, regardless of the biological entities they are assigned to. In October 2020, there were 44264 of these terms in the complete GO tree. GO terms are organized into three independent domains: molecular

function, biological process, and cellular component (Ashburner *et al.*, 2000). In each domain, terms are linked into a directed acyclic graph, an excerpt of which is shown in Figure 4.1. In the GO graph, a parent term can have one or more child (e.g., the root node “biological process” is the parent of the children GO:0009987 and GO:0008152), and children can have multiple parents (e.g., the most specific term “translation” has GO:0043043, GO:0034645, and GO:0044267 as parents).

Whereas this highly branched graph structure of GO allows flexible annotation at various levels of detail, it also creates problems when the results from one data set are compared to those of another data set. Indeed, even though two terms may be closely linked in the GO tree and are therefore highly similar (e.g., as parent and child terms or as sibling terms), the typically employed exact term matching will treat these terms as wholly unrelated, as the actual GO terms (and their accession numbers) are not identical. This problem is illustrated in a study by (Sajulga *et al.*, 2020), where a multisample data set was analyzed using several metaproteomics tools. The resulting GO terms were then compared using exact matching. The overlap between the result sets was quantified using the Jaccard index and was found to be low. As previously explained, this low similarity is likely the result of the limitations of the exact term matching approach.

There is thus a clear need for a more sophisticated GO term comparison that takes into account the existing relationships in the full GO tree. However, most existing tools that provide such comparison are based on enrichment analyses (Huang *et al.*, 2009; Waardenberg *et al.*, 2015; Fruzangohar *et al.*, 2013). In such analyses, a list of genes is mapped to GO terms, which are then analyzed for enriched biological phenomena. As a result, to the best of our knowledge, no tools allow the direct comparison of large functional data sets against each other, nor are these able to provide metrics to determine how functionally similar data sets are.

We therefore present MegaGO, a tool for comparing the functional similarity between large sets of GO terms. MegaGO calculates a pairwise similarity score between multiple sets of GO terms for each of the three GO domains and can do so in seconds, even on platforms with limited computational capabilities.



**Figure 4.1:** Excerpt of the biological process domain of the Gene Ontology showing all parent terms up to the root for “translation” (GO:0006412). The root GO term “biological process” (GO:0008150) has multiple children. The most specific term “translation”, in contrast, has multiple parents. When comparing the two terms GO:0044267 and GO:0034645 (portrayed in light red), we find two different lowest common ancestors: GO:0044249 and GO:1901576 (dark red). Only one of these, however, can be the most informative common ancestor (MICA), that is, the common ancestor with the highest information content for the terms in light red. Because an IC of 1.52 is larger than 1.48, the GO:0044249 is the MICA. The terms GO:0043604 and GO:0006518 (in light blue) are more similar than the two terms we previously described and have only one lowest common ancestor, which is also automatically the MICA for these terms: GO:0043603 (in dark blue). IC, information content; \*, most informative common ancestor.



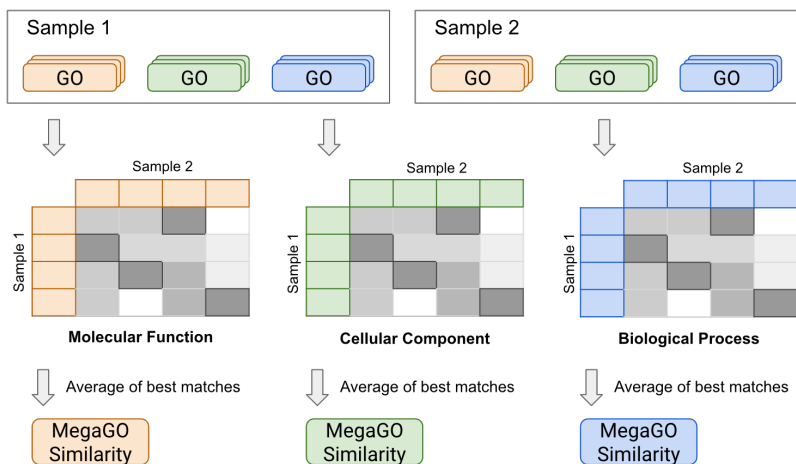
## 4.2 Implementation

To measure the similarity between sets of GO terms, we first need to measure the similarity of two individual terms. We compare two terms using the Lin semantic similarity metric, which can take on a value between 0 and 1 (Supplementary Formula 1a). The Lin semantic similarity is based on the ratio of the information content of the most informative common ancestor (MICA) to the average of the terms' individual information content.

The information content (Supplementary Formula 1b) is computed by estimating the terms' probability of occurrence (Supplementary Formula 1c), including that of all of their children. Term frequencies are estimated based on the manually curated SwissProt database (The UniProt Consortium, 2019). As a result, a high-level GO term such as "biological process" (through its many direct or indirect child terms) will be present in all data sets and thus carries little information. A more specific term such as "translation" (or any of its potential child terms) will occur less frequently and thus will be more informative (Figure 4.1). To finally calculate the similarity of two terms, we compare their information content with that of their shared ancestor that has the highest information content, the MICA. If the information content of the MICA is similar to the terms' individual information content, then the terms are deemed to be similar. The dissimilar terms "peptide biosynthetic process" and "cellular macromolecule biosynthetic" are situated further from their MICA "cellular biosynthetic process" than the similar terms "amide biosynthetic process" and "peptide metabolic process" with their respective MICA "cellular amide metabolic process" (Figure 4.1).

MegaGO, however, can compare not only two terms but also sets of GO terms. More specifically, two sets of GO terms can be compared via the web application, but an unlimited number of sets can be compared via the command line tool. Note that in these sets, duplicate GO terms will be removed so that each GO term will be equally important, regardless of how often it is provided by the user. To compare the sets of GO terms, pairwise term similarities are aggregated using the Best Matching Average (BMA, Supplementary Formula 2) (Schlicker *et al.*, 2006). For each GO term in the first input data set, the BMA finds the GO term with the highest Lin semantic similarity in the second data set and averages the values of these best matches. Moreover, MegaGO calculates the

similarity for each of the three domains of the gene ontology (molecular function, biological process, and cellular component), as GO terms from distinct domains do not share parent terms. The general overview of MegaGO is shown in Figure 4.2.



**Figure 4.2:** Overview of MegaGO workflow. The Gene Ontology (GO) terms of each sample set are separated into three GO domains: molecular function, cellular component, and biological process. Each term of each sample set is compared to every term in the other set that is from the same domain. The match with highest similarity for each term is then selected, and the average across all of these best matches is calculated.

MegaGO is implemented in Python, is installable as a Python package from PyPi, and can easily be invoked from the command line. The GOATOOLS (Klopfenstein *et al.*, 2018) library is used to read and process the Gene Ontology and to compute the most informative common ancestor of two GO terms, which are both required to compute the information content value (Supplementary Formula 1,  $p(\text{go})$ ). GO term counts are recomputed with every update of SwissProt, and a new release is automatically published bimonthly to PyPi, which includes the new data set. Automated testing via GitHub Actions is in place to ensure correctness and reproducibility of the code. In addition, we also developed a user-friendly and easily accessible web application

that is available on <https://megago.ugent.be>. The backend of the web application is developed with the Flask web framework for Python, and the frontend uses Vue. Our web application has been tested on Chromium-based browsers (Chrome, Edge, and Opera) as well as Mozilla Firefox and Safari. The MegaGO application is also available as a Docker-container on Docker Hub (<https://hub.docker.com/repository/docker/pverscha/mega-go>) and can be started with a single click and without additional configuration requirements. Our Docker container is automatically updated at every change to the underlying MegaGO code. All code is freely available under the permissive open source MIT license on <https://github.com/MEGA-GO/>. Documentation for our Python script can be found on our Web site: <https://megago.ugent.be/help/cli>. A guide on how to use the web application is also available: <https://megago.ugent.be/help/web>.

MegaGO is cross-platform and runs on Windows, macOS, and Linux systems. The system requirements are at least 4 GiB of memory and support for either Python 3.6 (or above) or the Docker runtime.

### 4.3 Validation

To validate MegaGO, we reprocessed the functional data from (Easterly *et al.*, 2019). This data set consists of 12 paired oral microbiome samples that were cultivated in bioreactors. Each sample was treated with and without sucrose pulsing, hereafter named ws and ns samples, respectively. Each sample contains mass-spectrometry-based proteomics measurements, and all samples were annotated with 1718 GO terms on average. We calculated the pairwise similarity for each of the 300 sample combinations, which took less than 1 min for a single sample pair on the web version of MegaGO. This resulted in a MegaGO similarity score for each of the three GO domains for each sample combination. These similarities were then hierarchically clustered and visualized in a heatmap. All data and intermediate steps of our data analysis are available at <https://github.com/MEGA-GO/manuscript-data-analysis/> and can be reproduced with the command line tool using the `-heatmap` option.

In the heatmap (Figure 4.3), we can observe that the two sample groups cluster together, except for 730ns and 733ns that are clustered in the ws sample group. These two samples were also identified as outliers in

(Easterly *et al.*, 2019) and 733ns was originally also identified as both a taxonomic and functional outlier in (Rudney *et al.*, 2015). Similar results can be observed for the GO domain “molecular function” (Supplementary Figure 1). The MegaGO similarity-based clustering of “cellular component” GO terms (Supplementary Figure 2) has two additional samples clustered outside of their treatment group: 852ws in the ns cluster and 861ns in the ws group. Again, these patterns can also be found in previous analyses: 852ws is placed in direct proximity of the ns samples in the principal component analysis (PCA) of the HOMINGS analysis by Rudney *et al.*, and 861ns is closest to 730ns and 733ns in PCA of Rudney *et al.*’s taxonomic analysis. Interestingly, subjects 730 and 852 were the only ones without active carious lesions, which could cause their divergence in the similarity analyses.

Results produced by MegaGO are thus in close agreement with prior analyses of the same data, showing that MegaGO offers a valid and very fast approach for comparing the functional composition of samples.

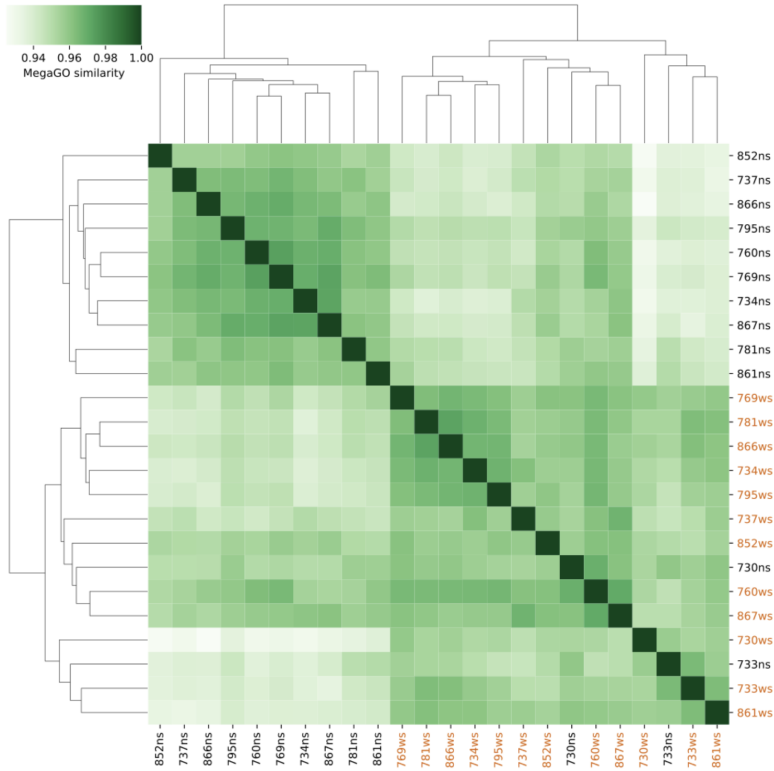
## 4.4 Conclusions

MegaGO enables the comparison of large sets of GO terms, allowing users to efficiently evaluate multiomics data sets containing thousands of terms. MegaGO separately calculates a similarity for each of the three GO domains (biological process, molecular function, and cellular component). In the current version of MegaGO, quantitative data are not taken into account, thus giving each GO term identical importance in the data set.

MegaGO is compatible with any upstream tool that can provide GO term lists for a data set. Moreover, MegaGO allows the comparison of functional annotations derived from DNA-, RNA-, and protein-based methods as well as combinations thereof.

## 4.5 Acknowledgements

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**Figure 4.3:** Hierarchically clustered heatmap comparing MegaGO similarities for the GO domain “biological process” for each of the samples from (Easterly *et al.*, 2019). Samples that are treated with sucrose pulsing are labeled as “ws” and are displayed in orange.

Berlin, Germany) and Stephan Fuchs (Robert Koch Institute, Berlin, Germany) for their support. P.V., T.V.D.B., L.M., and B.M. acknowledge the Research Foundation - Flanders (FWO) (grants 1164420N, 1S90918N, G042518N, and 12I5220N). L.M. also acknowledges support from the European Union's Horizon 2020 Programme under grant agreement 823839 (H2020-INFRAIA-2018-1). H.S. and B.Y.R. acknowledge support by Deutsche Forschungsgemeinschaft (DFG; grant numbers RE3474/5-1 and RE3474/2-2) and the BMBF-funded de.NBI Cloud within the German Network for Bioinformatics Infrastructure (de.NBI; 031A537B, 031A533A, 031A538A, 031A533B, 031A535A, 031A537C, 031A534A, and 031A532B).

## Chapter 5

# Unipept CLI 2.0: adding support for visualisations and functional annotations

**Abstract** Unipept (Mesuere *et al.*, 2012) is a collection of tools developed for fast metaproteomics data analysis. The Unipept ecosystem consists of a web application, an application programming interface (API) as a web service (Mesuere *et al.*, 2016) and a command-line interface (CLI) (Mesuere *et al.*, 2018). The key strengths of Unipept are its speed, its ease-of-use and the extensive use of interactive data visualization in the analysis results. The Unipept database is derived from the UniProt (UniProt, 2019) KB and consists of tryptic peptides linked with taxonomic and functional annotations. Unipept initially launched with support for taxonomic analysis of metaproteomics data in 2012. Version 4.0 (Gurdeep Singh *et al.*, 2019) of the Unipept web application was launched in November 2018 and extended the web interface with support for functional annotations such as Gene Ontology (GO) terms (Ashburner *et al.*, 2000), Enzyme Commission (EC) numbers (Webb, 1992) and InterPro entries (Hunter *et al.*, 2009).

## 5.1 Introduction

Unipept (Mesuere *et al.*, 2012) is a collection of tools developed for fast metaproteomics data analysis. The Unipept ecosystem consists of a web application, an application programming interface (API) as a web service (Mesuere *et al.*, 2016) and a command-line interface (CLI) (Mesuere *et al.*, 2018). The key strengths of Unipept are its speed, its ease-of-use and the extensive use of interactive data visualization in the analysis results. The Unipept database is derived from the UniProt (The UniProt Consortium, 2019) KB and consists of tryptic peptides linked with taxonomic and functional annotations. Unipept initially launched with support for taxonomic analysis of metaproteomics data in 2012. Version 4.0 (Gurdeep Singh *et al.*, 2019) of the Unipept web application was launched in November 2018 and extended the web interface with support for functional annotations such as Gene Ontology (GO) terms (Ashburner *et al.*, 2000), Enzyme Commission (EC) numbers (Webb, 1992) and InterPro entries (Hunter *et al.*, 2009).

The GO terms are organized into three different domains: ‘cellular components’, ‘molecular functions’ and ‘biological processes’. Every GO-term is associated with exactly one domain and consists of a name, an identifier and an exact definition. The EC numbers can be used to classify enzymes, based on the chemical reactions that they catalyze. Every EC number consists of four numbers, separated by a dot, yielding a hierarchical classification system with progressively finer enzyme classifications. InterPro is a database that consists of predictive models collected from external databases that can be classified into five different categories. More information about functional annotation in metaproteomics can be found in the study by (Schiebenhoefer *et al.*, 2019).

For each input peptide, Unipept finds all functional annotations associated with all of the UniProt entries in which the peptide occurs. All found functions are listed in order of decreasing number of peptides associated with this function.

In this article, we present several new additions to the Unipept API and CLI which allow third-party applications [such as Galaxy-P (Jagtap *et al.*, 2015)] to integrate the new functional analysis capabilities provided by Unipept.



## 5.2 Materials and methods

The Unipept API is a high-performance web service that responds in a textual format (JSON) to HTTP-requests from other applications or tools and allows to integrate the services provided by Unipept into other workflows. Unipept's CLI is a Ruby-based application and high-level entry point which allows users to actively query Unipept's database. Compared to the API, users do not need to compile API-requests manually but can rely on the CLI to automatically do so in a parallelized way. In addition, it supports multiple input and output formats such as FASTA and CSV.

The Unipept database and web application were recently expanded to include GO terms, EC numbers and InterPro entries. These new annotations are now also available from newly developed API-endpoints and CLI-functions, providing structured access to this functional information.

Most of the newly developed endpoints support batch retrieval of information for a list of peptides. In this case, the API returns a list of objects where each object in the response corresponds with information associated with one of the input peptides. Every API-endpoint is accompanied by an identically named CLI-function, which provides the user with the ability to import data from or export data to various specifically formatted files. In addition, version 2 of the Unipept CLI introduces the ability to produce interactive visualizations directly from the command line.

Among other information, the Unipept tryptic peptide analysis lists functional annotations associated with a given tryptic peptide. These data are aggregated because a peptide can occur in multiple proteins that each can have multiple functional annotations. For each annotation, we also return the amount of underlying proteins that match with this specific annotation.

Some applications require all known information for a list of tryptic peptides. The 'pept2func' function is a combination of the preceding three endpoints and returns all functional annotations associated with the given tryptic peptide. 'peptinfo' on the other hand, returns all the available information for one or more tryptic peptides. All functional annotations for this peptide are part of the response, as well as the lowest common ancestor for this peptide. Both functions also support

splitting the GO terms and InterPro entries over the respective domains, and naming information can optionally be retrieved.

The ‘taxa2tree’ function constructs a tree from a list of NCBI taxon ids. This tree is an aggregation of the lineages that correspond with the given taxa and can be exported as three distinct output formats: JSON, HTML and as a URL. The HTML and URL representation of a taxonomic tree both provide three interactive data visualizations, albeit with different possibilities. A generated HTML-string first needs to be stored in a file before it can be rendered by a browser and cannot be easily shared with other people but is easily editable. A URL on the other hand is simply a shareable link to an online service that hosts all interactive visualizations.

## 5.3 Conclusion

Version 2.0 of the Unipept API and CLI is a significant update that provides fast and easy access to the powerful analysis pipeline of Unipept. In addition to the existing taxonomic analysis, it now features multiple functional annotations which will enable users to gain new insights into complex ecosystems. These new features can easily be integrated into third-party tools such as the MetaProteome Analyzer (Muth *et al.*, 2018). Galaxy-P, a highly used workflow integration system, is already successfully making use of the novel analysis functions that are introduced with this new release.

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