New targeted approaches for the quantification of data-independent acquisition mass spectrometry
 Christine Stadelmann ${ }^{4,5}$, Alexey I. Nesvizhskiii3,5, Manuela Schmidt ${ }^{2,5}$, Lukas Reiter ${ }^{1,5,}$, and David Gomez-Varela ${ }^{25,{ }^{*}}$
${ }^{1}$ Biognosys AG, Schlieren, Switzerland
${ }^{2}$ Somatosensory Signaling and Systems Biology Research Group, Max Planck Institute of Experimental Medicine, Goettingen, Germany
${ }^{3}$ Department of Computational Medicine and Bioinformatics, University of Michigan, Ann Arbor, Michigan,
${ }^{4}$ Department of Pathology, University of Michigan, Ann Arbor, Michigan, USA
${ }^{5}$ Institute of Neuropathology, University Medical Center, Goettingen, Germany
*Correspondence: David Gómez Varela, Hermann-Rein Strasse 3, D-37075, Goettingen, Germany. Email: gomez@em.mpg.de, Fax: +49 (0) 5513899573
*These authors contribute equally to this work.


#### Abstract

Abstraçt $\longrightarrow$ The use of data-independent acquisition (DIA) approaches for the reproducible and precise quantification of complex protein samples has increased in the last years. The protein information arising from DIA analysis is stored in digital protein maps (DIA-maps) that can be interrogated in a targeted way by using ad hoc or publically-available peptide spectral libraries generated on the same sample species as for the generation of the DIA maps. The restricted availability of certain difficult-to-obtain human tissues (i.e. brain) together with the caveats of using spectral libraries generated under variable experimental conditions limits the potentiat of DIA. Therefore, DIA workflows would benefit from high-quality and extended spectral libraries that could be generated without the need of using valuable samples for library production. We describe here two new targeted approaches, using either classical data-dependent (DDA) repositories (not specifically built for DIA) or ad hoc mouse spectral libraries, which enable the profiling of human brain DIA dataset. The comparison of our results to both the most extended publically-available human spectral library and to a state-of-the-art untargeted method supports the use of these new strategies to improve future DIA profiling efforts.


This is the author manuscript accepted for publication and has undergone full peer review but has not been through the copyediting, typesetting, pagination and proofreading process, which may lead to differences between this version and the Version of Record. Please cite this article as doi: 10.1002/pmic. 201700021.

This article is protected by copyright. All rights reserved.

## Significance of the study

A lot of interest has been generated around DIA methods applied to discovery-based proteomics, as they promise to overcome the well-known stochastic nature of classical DDA schemes. However, new analytical solutions are needed in order to increase the analyzed landscape of the DIA maps.

In this study we present a targeted analytical strategy to increase the quantitative information recovered from human DIA maps by using either the extended sources stored in classical DDA repositories or ad hoc mouse spectral libraries.

To the best of our knowledge this is the first demonstration of the possibility to unlock years of DDA research for the analysis of DIA datasets, which we expect will have an important impactin the growing field of DIA proteomics.

The changes in the proteome underlying the regulation of biological processes is of major interest in biomedical research and relies on the ability of proteomic methods to generate quantitative, reproducible and comprehensive datasets. To date, DDA mass spectrometry methods have been extensively used. In DDA, a subset of peptide precursors is selected from a survey scan (MS1) for subsequent fragmentation and acquisition of a fragment-ion speetrum (MS2). Their identification together with the quantitative information results in the discovery of differences in protein expression. However, the selection of peptide precursors is stochastic favoring the collection of high-abundant/intense peptides. As a result, roughly $27 \%$ of all detectable peptides can be identified in a single run [1, 2], and $30 \%$ of the selected peptides can vary between replicates [3]. As an alternative approach, pioneer work [4-8] showed the potential of DIA methods where all peptide ions (in a given $\mathrm{m} / \mathrm{z}$ range and above the detection limit of the LC-MS instrument) of a complex sample are fragmented independently of their intensity, theoretically enabling the identification and quantification of all peptide precursors. However, DIA-MS2 spectra stored in the form of DIAmaps are composed by fragmented ions belonging to different peptide precursors. This chimeric nature of the MS2 spectra is a bottleneck for identification using classical database search strategies.

The most widely used analytical solution of DIA methods is the use of reference spectral libraries for the targeted extraction of quantitative information of the peptides included in these ibraries [9] using tools such as Spectronaut, OpenSWATH, Skyline or PeakView. While this type of targeted approaches reaches excellent reproducibility [10], the analysis is restricted to a subset of the comprehensive peptide information stored in DIAmaps because reference spectral libraries are generated by DDA proteomics on the same sample species. Public repositories containing species-specific spectral libraries (i.e. Peptide Atlas) aim to solve this bottleneck, however the limited availability of certain tissue samples together-with the caveats of using spectral libraries generated under variable experimental conditions [11, 12] prevents DIA to achieve its full potential.

Based on the correspondence between the peptide fragmentation pattern and the peptide sequence [13], we sought to improve the targeted analysis of human DIA-maps by developing two novel and complementary strategies. On the one hand, we used classical DDA repositories as means of human reference spectral libraries. On the other hand, we

This article is protected by copyright. All rights reserved.
generated ad hoc spectral libraries using mouse tissues, a widely available and closely related animal model. Our results demonstrate our approaches having an improved performance in analyzing human DIA-maps when compared to the most extended panhuman spectral library [14] and to a recent untargeted approach [15].

First, we isolated regions of spinal cord (SC) and prefrontal cortex (PFC) from human and mouse (Fig. 1A). Our selection was guided by the important roles of these tissues in human pathophysiology and their low availability compared to other frequently used human matrices (e.g. plasma). Following, we constructed extended mouse and human ad hoc libraries (Mil and HI, respectively) using a combined strategy. In order to reach the most comprehensive coverage possible of the expressed proteome we use multiple injections and long chromatography gradients, as well as three tissue fractions for each of the tissues: whole cell, membrane- and cytosol-enriched (Supporting Information Fig. 1 and Fig. 2). Finally, we created a mouse-derived human library (MdHI) by selecting only peptides displaying $100 \%$ identity with their human counterparts as annotated in Uniprot (of the 43,956 mouse peptides 26,953 overlap with human peptides).

As an additional strategy, we aimed to construct high-quality spectral libraries using the currently most extended datasets from classical DDA repositories of mouse and human cortex [16-18]. This relies on normalized retention times and on the ability to predict accurate retention times (RT) for peptides detected in different experimental conditions. For that, we applied a high-precision iRT algorithm [11] to all peptides from the mentioned datasets that resulted in two DDA-repository (dd) spectral libraries: a dd mouse-derived human library (ddMdHI) and a dd human library (ddHI).

Comparison of the four generated libraries with the currently most-extended panhuman spectral library ( PHI ) [14] revealed a significant amount of newly incorporated proteins (Fig. 1B and Supporting Information Table 1). A closer look revealed that 223 proteins of the HI are not present in the PHI but are included in the three other libraries of which a significant amount are involved in physiological functions assumed by PFC and SC (e.g. transmission of nerve impulse; Supporting Information Fig. 3).

Next, we evaluated the ability of the generated libraries to profile differences in protein expression in human tissues. To this end, we obtained DIA-maps from three biological replicates of whole cell lysates of human SC and PFC (Fig. 1). Our results indicate that HI profiled the highest number of unique peptides and proteins in both tissues and in all of the three biological replicates (Fig. 2, Supporting Information Table 2 and Supporting Information Fig. 4A-C), including also the highest number of peptides representing each protein (Supporting Information Fig. 4D). Importantly, ddHI, ddMdHI and MdHI outperformed PHI (114\% more peptides for ddHI in PFC; Fig. 2 and Supporting Information Table 2) with comparable coefficients of variation (Supporting Information Fig. 5). The improvements shown here with inter-species resources (mouse) differ conceptually from previously published DDA database search strategies [19] as our approach does not rely on modifying the size of the database search space or on modifying the DIA map search space in order to influence the probability of positive peptide identification.

Our data partially confirm but also demonstrate the potential of our strategy to overcome well-known limitations when aiming to use spectral data obtained in different experimental conditions (i.e. different mass spectrometers which tend to produce different charge states of

This article is protected by copyright. All rights reserved.
a given peptide). A closer look at the results show that the MdHI (generated using the same QExactive mass spectrometer as for the DIA maps) was able to extract complete quantitative information for 893 proteins more than the PHI (generated using a TripleTOF $5600+$ mass spectrometer) in human PFC (Supporting Information Table 2), although 506 of those proteins were also represented in the PHI. A deeper analysis revealed that $56 \%$ of these proteins were represented by different peptides in the PHI spectral library, whereas in the remaining $44 \%$ different charge states, retention time and fragment ion selections were apparent. These differences go in line with the reported differences in spectral transferability between different instruments [20]. On the other hand we believe that the tissue similarity plays an important role: ddHI that is derived from the analysis of human frontal cortex (Supporting Information Table 1) is able to extract information of 32 more proteins than MdHI (Supporting Information Table 2) although the latter library was produced using exactly the same instrumentation as for the human DIA datasets. Further, ddHI extracted 925 proteins more than PHI (Supporting Information Table 2) that did not include any brain tissue for its generation [14]. Th summary, our results show that although ad-hoc in-house spectral libraries from the same species as the DIA datasets and obtained using the same instrumentation continue as the most efficient choice, there is a big potential of applying high-precision iRT algorithms [11] to extended datasets obtained from classical DDA repositories (not specifically built for DIA).

Following, we evaluated whether the peptide assays included in all these libraries harbor not only proteotypic but also quantitative properties [21]. As shown in Figure 3A, the differences in abundance detected by ddHI, in comparison to HI showed remarkable consistency - 525 proteins were commonly identified with ddHI and HI to be differentially expressed, including important synaptic proteins (Fig. 3B). Comparable results were obtained for ddMdHI and MdHI (Supporting Information Fig. 6).

Recent computational efforts have been developed to analyze DIA datasets without the need for spectral libraries [15, 22, 23] - referred to as untargeted approaches. Our data shows that library-based targeted extraction using $\mathrm{HI}, \mathrm{MdHI}$ and ddHI outperformed the untargeted strategy of DIAUmpire in terms of number of extracted peptides and proteins (Fig. 2, Supporting Information Table 2, Supporting Information Fig. 4 and Fig. 5). An alternative for the analysis of our human DIA datasets without relying in the use of ad hoc HI or MdHI would be to use DIA-Umpire to generate a complementary library and combine it with the identifications when using the publically-available PHI. Our results indicate that the identifications reached by ddHI or MdHI alone or MdHI grouped with ddMdHI still outperformed a combination of DIA-Umpire and PHI (Fig. 2, Supporting Information Table 2 and Supporting Information Fig. 4).

The detection improvements over DIA-Umpire were comparable to or even bigger than the ones reached by latest untargeted solutions when similar conditions were compared: While MdHI alone and in combination with ddMdHI showed an improvement of $29 \%$ and $40 \%$, respectively (Supporting Information Table 2), MSSPLIT-DIA [22] showed an approximate $25 \%$ improvement using a spectral library generated from the same sample. On the other hand, Group-DIA [23] did not show any improvement in peptide identifications when 3 DIA runs were considered (a similar number of runs as analyzed to construct the DIA maps in our study).

Interestingly, 204 of the peptides detected by DIA-Umpire were not present in any of the five other libraries. Further analysis showed that the intensities of the fragment ions exclusively profiled by DIA-Umpire were significantly lower in comparison to the ones commonly detected with the other five libraries (two sample $t$-test, two-tailed, $\mathrm{P}<0.0001$; similarity of the variance was confirmed using F test, $\mathrm{P}=0.13$; Fig. 3C). These results suggest that future integrative strategies would benefit from the combination of targeted and untargeted approaches in order to increase the analysis-depth of DIA-datasets.

DIA methods are in a prime position for proteomic-based biomarker discovery efforts. Several strategies are under development towards the goal to improve the amount of reliable peptide identifications from DIA maps, including computational [15, 22, 23] and technological efforts [24]. The novel strategies presented here add new possibilities to these efforts that can be easily implemented in future studies, unlocking extended classical DDA repositories [25] for comprehensive profiling of human samples by DIA.

EXPERIMENTAL PROCEDURES

## Human and mouse tissue

Native, unfixed, shap frozen CNS tissue of prefrontal cortex (PFC) and spinal cord (SC) of three non-neurological patients was obtained from the archives of the Institute of Neuropathology, University Medical Center, Göttingen (UMG), Germany. Post mortem tissue sampled for diagnostic purposes was used for the present study in conformity with the rules and regulations of the ethics committee of the UMG, reference number $6 / 5 / 16$. Similar regions from three male 12-14 weeks old wild type C57BI/6J mice were obtained. No randomization or blinding was used in this study. All mouse experiments are approved by the IACUC of the Max Planck Institute of Experimental Medicine (MPIEM).

## Sample preparation for mass spectrometry acquisition

The frozen tissue was homogenized with help of a glass/Teflon homogenizer in 4\% SDS lysis buffer ( $4 \%$ SDS in 100 mM Tris, 10 mM DTT, $5 \%$ glycerol, complete protease inhibitor cocktail (Roche), pH 7.5 and by shearing with a 25G needle. The homogenate was incubated for 10 minutes at $70^{\circ} \mathrm{C}$, followed by centrifugation at $10,000 \mathrm{xg}$ for 5 minutes for removal of cell debris at room temperature (note: all centrifugation steps in this study were done al room temperature, except otherwise mentioned). The supernatant equals the whole cell lysate. Following, acetone precipitation of the proteins was done by addition of $5 x$ volume pre-cooled acetone (Roth) and incubation for 2 hours at $-20^{\circ} \mathrm{C}$. The precipitated proteins were centrifuged at $14,000 \mathrm{xg}$ for 30 minutes, washed with ice-cold $80 \%$ Ethanol (AppliedChem) and centrifuged again at $14,000 \mathrm{xg}$ for 30 minutes. The air-dried proteins were resuspended under constant agitation in $2 \%$ SDS lysis buffer. For generation of the membrane and cytosolic fractions, the whole cell lysate was subjected to ultracentrifugation at $100,000 \mathrm{xg}, 18^{\circ} \mathrm{C}$ for 1 hour (TLA 100.3 rotor, Beckman Coulter). The pellet was dissolved in 4\% SDS lysis buffer. The supernatant (cytosol) and dissolved pellet (membrane fraction) were subjected to acetone precipitation as described above.

The samples in $2 \%$ SDS were prepared for mass spectrometric acquisition using the FASP protocol with 30k Vivacon 500 spinfilters[26] (Sartorius). The peptides were desalted using

C18 MacroSpin columns from The Nest Group according to manufacturer's instructions. After drying, the peptides were resuspended in $1 \%$ ACN and $0.1 \%$ formic acid. The Biognosys' HRM Calibration Kit, was added to all of the samples according to manufacturer's instructions (required for the DIA analysis using Biognosys' Spectronaut). All the three biological replicates for each of the four tissues (two in human and two in mouse) were processed in parallel.

Mass spectrometric acquisition
$2 \mu \mathrm{~g}$ of the samples was analyzed on a self-made analytical column ( $75 \mu \mathrm{~m} \times 30 \mathrm{~cm}$ ) packed with $3 \mu \mathrm{~m}$ Magic C18AQ medium (Bruker) at $50^{\circ} \mathrm{C}$, using an Easy-nLC connected to a Q Exactive mass spectrometer (Thermo Scientific). The peptides were separated by a 2 hours segmented gradient from 0 to $9 \%$ ACN in 3 minutes, next to $28 \%$ ACN in 102 minutes, to $36 \%$ ACN in 12 minutes and to $44 \%$ in 3 minutes, with $0.1 \%$ formic acid at $300 \mathrm{nl} /$ minute. Then a linear increase followed to $90 \%$ ACN in 2 minutes and $90 \%$ ACN was run for 8 minutes. For DDA acquisition, the "fast" method from Kelstrup [27] was used with the alteration described in Bruderer et al. [28]. In order to produce the most extended spectral libraries possible we have analyzed 10 independent samples for each of the tissues.

The HRM DIA method was performed using one technical replicate for each of the three biological replicates, and as described in Bruderer et al. with the following modification: the survey scan was recorded at 70,000 resolution in order to reach superior data quality through the introduction of MS1 based scores. The raw mass spectrometric DIA data and the spectrallibraries were stored at the public repository PeptideAtlas (http://www.peptideatlas. org, No. PASS00782, the username is PASS00782 and the password is NN4585e).

## MaxQuant data analysis

The DDA spectra were processed with the MaxQuant Version 1.5.1.2 analysis software [29]. The spectra were searched by the Andromeda search engine [30]. The minimal peptide length was set to 6 . Search criteria included carbamidomethylation of cysteine as a fixed modification, oxidation of methionine and acetyl (protein N -terminus) as variable modifications. The mass tolerance for the precursor was 4.5 ppm and for the fragment ions was 20 ppm . The human DDA files were searched against a human UniProtKB/Swiss-Prot fasta database (state 11.12.2014, 20,199 entries), the contaminants of MaxQuant and the Biognosys iRT peptide sequences [31] (11 entries). The murine DDA files were searched against a mouse UniProtKB/Swiss-Prot fasta database (state 11.12.2014, 16,697 entries), the contaminants of MaxQuant and the iRT peptide sequences.

## Spectral library generation

This article is protected by copyright. All rights reserved.

To generate the spectral libraries, DDA spectra were analyzed as described above and a spectral library was generated using the spectral library generation functionality of Spectronaut (details can be found in the Supporting Information).

## Spectronaut analysis

The DIA data were analyzed with Spectronaut 8, a mass spectrometer vendor independent software from Biognosys following default settings (details can be found in the Supporting Information).

DIA-Umpire analysis
Additionally, the DIA data was analyzed using the DIA-Umpire workflow as described in the DIA-Umpire manuscript [15] (details can be found in the Supporting Information).

```
pmic201700021-sup-0001-Fig1.jpg
pmic201700021-sup-0002-Fig2.jpg
pmic201700021-sup-0003-Fig3.jpg
pmic201700021-sup-0004-Fig4.jpg
pmic201700021-sup-0005-Fig5.jpg
pmic201700021-sup-0006-Fig6.jpg
pmic201700021-sup-0007-Table1.xlsx
pmic201700021-sup-0008-Table2.xlsx
pmic201700021-sup-0009-text.docx
ACKNOWLEDGMENTS
This work was supported by grants of the Deutsche Forschungsgemeinschaft (DFG; GO 2481/3-1 to D.G.V., SCHM 2533/2-1 to M.S.), a MPI PhD fellowship (to J.S.), the US National Institutes of Health grants NIH R01-GM-094231 (to A.I.N), the Research Program (Forschungsförderungsprogramm) of the Faculty of Medicine, Georg-August-University, Göttingen and the Hertie Foundation (Gemeinnützige Hertie-Stiftung) (to C.S. and A.B).
```


## AUTHOR CONTRIBUTIONS

D.G.V. conceived and supervised the project. D.G.V., R.B. and C.S. designed experiments. J.S., M.S., A.B.F. performed tissue extraction and preparation. R.B. and L.R. acquired and analyzed mass spectrometry data. C.C.T. and A.I.N. performed the DIA-Umpire analysis. D.G.V. wrote the manuscript with input from all authors.

## COMPETING FINANCIAL INTERESTS

This article is protected by copyright. All rights reserved.

The authors R.B. and L.R. are employees of Biognosys AG (Switzerland). Spectronaut is a trademark of Biognosys AG. DGV received a research award from Biognosys AG that was not used for the current study. The rest of the authors declare no competing financial interests.
References

[1] Michalskí, A., Cox, J., Mann, M., More than 100,000 detectable peptide species elute in single shotgun proteomics runs but the majority is inaccessible to data-dependent LC-MS/MS. Journal of proteome research 2011, 10, 1785-1793.
[2] Scheltema, R.A., Hauschild, J. P., Lange, O., Hornburg, D., et al., The Q Exactive HF, a Benchtop mass spectrometer with a pre-filter, high-performance quadrupole and an ultra-high-field Orbitrap analyzer. Molecular \& cellular proteomics : MCP 2014, 13, 3698-3708.
[3] Liu, H., Sadygov, R. G., Yates, J. R., 3rd, A model for random sampling and estimation of relative protein abundance in shotgun proteomics. Anal Chem 2004, 76, 4193-4201.
[4] Venable, J. D., Dong, M. Q., Wohlschlegel, J., Dillin, A., Yates, J. R., Automated approach for quantitative analysis of complex peptide mixtures from tandem mass spectra. Nature methods 2004, 1, 39-45.
[5] Silva, J.G. Gorenstein, M. V., Li, G. Z., Vissers, J. P., Geromanos, S. J., Absolute quantification of proteins by LCMSE: a virtue of parallel MS acquisition. Molecular \& cellular proteomics : MCP 2006, 5, 144-156.
[6] Panchaud, A., Scherl, A., Shaffer, S. A., von Haller, P. D., et al., Precursor acquisition independent from ion count: how to dive deeper into the proteomics ocean. Anal Chem 2009, 81, 6481-6488.
[7] Carvalho, P. C., Han, X., Xu, T., Cociorva, D., et al., XDIA: improving on the label-free dataindependent analysis. Bioinformatics 2010, 26, 847-848.
[8] Martins-de-Souza, D., Faca, V. M., Gozzo, F. C., DIA is not a new mass spectrometry acquisition method. Proteomics 2017.
[9] Schubert, O. T., Gillet, L. C., Collins, B. C., Navarro, P., et al., Building high-quality assay libraries for targeted analysis of SWATH MS data. Nature protocols 2015, 10, 426-441.
[10] Selevsek, N., Ghang, C. Y., Gillet, L. C., Navarro, P., et al., Reproducible and consistent quantification of the Saccharomyces cerevisiae proteome by SWATH-mass spectrometry. Molecular \& cellular proteomics: MCP 2015, 14, 739-749.
[11] Bruderer, R., Bernhardt, O. M., Gandhi, T., Reiter, L., High-precision iRT prediction in the targeted analysis of data-independent acquisition and its impact on identification and quantitation. Proteomics 2016.

This article is protected by copyright. All rights reserved.
[12] Wu, J. X., Song, X., Pascovici, D., Zaw, T., et al., SWATH Mass Spectrometry Performance Using Extended Peptide MS/MS Assay Libraries. Molecular \& cellular proteomics : MCP 2016, 15, 2501-2514.
[13] Zhang, Z., Prediction of low-energy collision-induced dissociation spectra of peptides. Anal Chem 2004, 76, 3908-3922.
[14] Rosenberger, G., Koh, C. C., Guo, T., Rost, H. L., et al., A repository of assays to quantify 10,000 human proteins by SWATH-MS. Scientific data 2014, 1, 140031.
[15] Tsou, C.C., Avtonomov, D., Larsen, B., Tucholska, M., et al., DIA-Umpire: comprehensive computational framework for data-independent acquisition proteomics. Nature methods 2015, 12, 258-264, 257 p following 264.
[16] Sharma, K., Schmitt, S., Bergner, C. G., Tyanova, S., et al., Cell type- and brain region-resolved mouse brain proteome. Nat Neurosci 2015, 18, 1819-1831.
[17] Wilhelm, M., Schlegl, J., Hahne, H., Moghaddas Gholami, A., et al., Mass-spectrometry-based draft of the human proteome. Nature 2014, 509, 582-587.
[18] Kim, M. S., Pinto, S. M., Getnet, D., Nirujogi, R. S., et al., A draft map of the human proteome. Nature 2014, 509, 575-581.
[19] Frank, A. M., Monroe, M. E., Shah, A. R., Carver, J. J., et al., Spectral archives: extending spectral libraries to analyze both identified and unidentified spectra. Nature methods 2011, 8, 587-59
[20] Toprak, U.H., Gillet, L. C., Maiolica, A., Navarro, P., et al., Conserved peptide fragmentation as a benchmarking tool for mass spectrometers and a discriminating feature for targeted proteomics. Molecular \& cellular proteomics : MCP 2014, 13, 2056-2071.
[21] Worboys, J. D., Sinclair, J., Yuan, Y., Jorgensen, C., Systematic evaluation of quantotypic peptides for targeted analysis of the human kinome. Nature methods 2014, 11, 1041-1044.
[22] Wang, J., Tucholska, M., Knight, J. D., Lambert, J. P., et al., MSPLIT-DIA: sensitive peptide identification for data-independent acquisition. Nature methods 2015.
[23] Li, Y., Zhong, C. Q., Xu, X., Cai, S., et al., Group-DIA: analyzing multiple data-independent acquisition mass spectrometry data files. Nature methods 2015.
[24] Egertson, J. D., Kuehn, A., Merrihew, G. E., Bateman, N. W., et al., Multiplexed MS/MS for improved data-independent acquisition. Nature methods 2013, 10, 744-746.
[25] Omenn, G. S., Lane, L., Lundberg, E. K., Beavis, R. C., et al., Metrics for the Human Proteome Project 2015: Progress on the Human Proteome and Guidelines for High-Confidence Protein Identification. Journal of proteome research 2015, 14, 3452-3460.
[26] Wisniewski, J. R., Zielinska, D. F., Mann, M., Comparison of ultrafiltration units for proteomic and N-glycoproteomic analysis by the filter-aided sample preparation method. Anal Biochem 2011, 410, 307-309.

This article is protected by copyright. All rights reserved.
[27] Kelstrup, C. D., Young, C., Lavallee, R., Nielsen, M. L., Olsen, J. V., Optimized fast and sensitive acquisition methods for shotgun proteomics on a quadrupole orbitrap mass spectrometer.
Journal of proteome research 2012, 11, 3487-3497.
[28] Bruderer, R., Bernhardt, O. M., Gandhi, T., Miladinovic, S. M., et al., Extending the limits of quantitative proteome profiling with data-independent acquisition and application to acetaminophen treated 3D liver microtissues. Molecular \& cellular proteomics : MCP 2015.
[29] Cox, J,Mann, M., MaxQuant enables high peptide identification rates, individualized p.p.b.range mass accuracies and proteome-wide protein quantification. Nature biotechnology 2008, 26, 1367-1372.
[30] Cox, J., Neuhauser, N., Michalski, A., Scheltema, R. A., et al., Andromeda: a peptide search engine integrated into the MaxQuant environment. Journal of proteome research 2011, 10, 17941805.
[31] Escher, C., Reiter, L., MacLean, B., Ossola, R., et al., Using iRT, a normalized retention time for more targeted measurement of peptides. Proteomics 2012, 12, 1111-1121.

## FIGURE LEGENDS



This article is protected by copyright. All rights reserved.

Figure 1. Experimental Workflow. (A) Ad hoc and DDA-repository derived (dd) human- (H) and mouse-derived (Md) peptide spectral libraries (HI, MdHI, ddHI and ddMdHI, respectively) were used for the quantification of DIA maps (human DIA maps) obtained from human PFC and SC. Their performance was compared to the pan-human peptide spectral library (PHI) and to DIA-Umpire (DIA-Ump). (B) Venn diagram showing the number of proteins in each of the five libraries


Figure 2. Performance of each library in human PFC and SC. (A) Number of peptides profiled by each of the spectral libraries in the human PFC DIA dataset and (B) in the human SC DIA dataset.


Figure 3. Quantification properties of each of the libraries. (A) Ratio of expression (hPFC/hSC) for 525 proteins calculated based on Hl and ddHI. Slope of regression 0.86 with $a R^{2}=0.91$. (B) Examples of expression differences between the two human tissues for the presynaptic protein Syntaxin 1B and the postsynaptic protein GluR2 upon use of HI and ddHI. (C) Distribution of the 204 fragment ion intensities identified uniquely by DIA-Umpire (red bars) or 204 randomly selected fragments shared with at least one of the five spectral libraries (blue bars) from the human PFC DIA dataset analysis.


