Mitochondrial phosphoproteomes are functionally specialized across tissues

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Mitochondria are essential organelles whose dysfunction causes human pathologies that often manifest in a tissue-specific manner. Accordingly, mitochondrial fitness depends on versatile proteomes specialized to meet diverse tissue-specific requirements. Increasing evidence suggests that phosphorylation may play an important role in regulating tissue-specific mitochondrial functions and pathophysiology. Building on recent advances in mass spectrometry (MS)-based proteomics, we here quantitatively profile mitochondrial tissue proteomes along with their matching phosphoproteomes. We isolated mitochondria from mouse heart, skeletal muscle, brown adipose tissue, kidney, liver, brain, and spleen by differential centrifugation followed by separation on Percoll gradients and performed high-resolution MS analysis of the proteomes and phosphoproteomes. This in-depth map substantially quantifies known and predicted mitochondrial proteins and provides a resource of core and tissue-specific mitochondrial proteins (mitophos.de). Predicting kinase substrate associations for different mitochondrial compartments indicates tissue-specific regulation at the phosphoproteome level. Illustrating the functional value of our resource, we reproduce mitochondrial phosphorylation events on dynamin-related protein 1 responsible for its mitochondrial recruitment and fission initiation and describe phosphorylation clusters on MIGA2 linked to mitochondrial fusion.

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Introduction

Mitochondria are double-membrane-bound organelles with an essential role in homeostasis of eukaryotic cells. They are often referred to as the “powerhouse of the cell” because of their prominent function in bioenergetics. Among many other processes, they are also involved in several biosynthetic processes such as balancing redox systems, the regulation of metabolic by-products like reactive oxygen species (ROS) (Spinelli & Haigis, 2018) and hold a central role in cell death (Bock & Tait, 2020). The function and stability of mitochondria depend on their intrinsic bioenergetics regulation and finely orchestrated interaction with the cellular microenvironment. Energy conversion via the oxidative phosphorylation system plays an essential role in harvesting energy from ingested nutrients. Moreover, the morphology of mitochondria within an eukaryotic cell is actively regulated by fusion and fission events which dynamically modulate their number, size, and localization (Liesa et al, 2009). Regulation of mitochondrial dynamics also affects the interplay of mitochondria with other cellular structures, such as the cytoskeleton for active regulation of their localization (Moore & Holzbaur, 2018), and organelles like the endoplasmic reticulum (ER) and lipid droplets to regulate many physiological processes such as energy metabolism and ion buffering. The mitochondria-associated membrane, which is the contact site of the outer mitochondrial membrane with the ER, comprises a unique set of proteins mediating this interaction and fine-tunes mitochondrial functions with the cellular microenvironment (Nunnari & Suomalainen, 2012; Kwak et al, 2020). Dysregulation of any of these intricate processes can lead to severe mitochondrial dysfunctions and diseases, including neurodegenerative diseases, cardiovascular disorders, myopathies, obesity, and cancers, which can manifest in a cell type- and tissue-specific manner (Suomalainen & Batterstby, 2018).

The fitness of mitochondria depends on the production and maintenance of functional and versatile proteomes specialized to carry out a variety of functions within the eukaryotic metabolism and meet diverse cellular and tissue-specific requirements (Kuznetsov et al, 2009). The mitochondrial proteome includes over a 1,000 proteins (see below), but only a small fraction of 13 proteins are encoded on the circular mitochondrial DNA molecule (Anderson et al, 1981). Thus, most of the mitochondrial proteins are encoded by the nuclear genome, synthesized outside of mitochondria and subsequently imported into the organelle, implying that mechanisms controlling mitochondrial protein quality (e.g., correct protein folding and import) are essential for health and integrity of mitochondria (Jadiya & Tomar, 2020). Furthermore, investigations of mitochondrial dynamics and functional plasticity have revealed regulatory roles for posttranslational modifications, including phosphorylation (Niemi & Pagliarini, 2021). Studies have shown that phosphorylation of several
mitochondrial proteins is involved in the regulation of central processes such as metabolic function, for instance, through phosphorylation of the E1-alpha subunit of PDH (Patel et al, 2014), mitophagy (Kolitsida et al, 2019), and fission (Cribbs & Strack, 2007; Taguchi et al, 2007; Ducommun et al, 2015; Toyama et al, 2016; Lewis et al, 2018). Thus, deregulation of protein phosphorylation might be an important underlying feature of mitochondrial physiology and pathophysiology. Currently, there is a significant knowledge gap of the mitochondrial variable proteomic composition and to what extent it is phosphorylated in a tissue-specific manner and how posttranslational regulation influences organelle function. A detailed understanding of the functional specialization of mitochondria at the protein and phosphorylation levels is needed to elucidate the contribution of mitochondria to health and disease.

Large-scale mass spectrometry (MS)-based quantitative proteomics studies from our and other groups have already shed light on the proteomic composition of mitochondria of various mammalian tissues and cell types, mostly highlighting that most of the proteins are shared between mitochondria of different tissues (Mootha et al, 2003; Forner et al, 2006; Pagliarini et al, 2008). The breadth and depth of such studies has been largely driven by technological advances in the field in combination with improvements of mitochondria isolation procedures, such as differential centrifugation (DC), DC in conjunction with ultracentrifugation on, for example, Percoll gradients, magnetic bead-assisted methods (MACS) (Kappler et al, 2016) or MitoTags (Bayraktar et al, 2019). Efforts in defining the mitochondrial proteome lead to databases like MitoCarta2.0 and IMPI (http://impi.mrc-mbu.cam.ac.uk/), both integrated in Mitominer4.0 (Smith & Robinson, 2019). A recent quantitative and high confidence proteome of human mitochondria identified 1,134 different proteins that vary over six orders in magnitude in abundance (Morgenstern et al, 2021). Similarly, efforts have been undertaken to map the mitochondrial phosphoproteome and identify dozens to hundreds of phosphorylation sites on mitochondrial proteins (listed in Table S1 [Schulenberg et al, 2003; Hopper et al, 2006; Lee et al, 2007; Feng et al, 2008; Lewandrowski et al, 2008; Aponte et al, 2009; Boja et al, 2009; Cui et al, 2010; Deng et al, 2010, 2011; Grimsrud et al, 2012; Bak et al, 2013; Zhao et al, 2011, 2014; Kruse & Højlund, 2017; Li et al, 2019; Guedouari et al, 2020; Sathe et al, 2021]). However, there has been a dramatic improvement in the technology of phosphoproteomics workflows during the last years, leading to the routine identification and quantification of thousands of phosphorylation sites in cell culture and in vivo systems (Humphrey et al, 2018; Bekker-Jensen et al, 2020) which had not been available in earlier studies. Furthermore, comparisons between mitochondrial phosphoproteomes have been difficult because only a single or a few tissues were analyzed. This complicates the combination and comparison of datasets across studies to obtain a clear view of mitochondrial diversity on proteome and phosphoproteome levels. Thus, a concerted effort is needed to systematically and quantitatively profile mitochondrial proteomes together with their matching phosphoproteomes from the same biological source. This would further help to investigate the dynamic composition of mitochondria and help identify the tissue-specific repertoire of mitochondria-resident kinases and phosphatases and their substrate associations.

Here, we performed a systematic analysis of the mitochondrial composition at the level of proteins, major functional entities, and phosphorylation in seven mouse tissues—brain, brown adipose tissue (BAT), heart, kidney, liver, skeletal muscle (SKM), and spleen. Our study employs state-of-the-art MS-based proteomics technology to systematically map divergent composition and phosphorylation of mitochondria between tissues and provides functionally valuable insights into their proteome and posttranslational regulations. This study contributes to our understanding of tissue-specific mitochondrial processes controlled by protein abundance and phosphorylation which is essential to manipulate these in health and disease. Our mitochondrial (phospho)proteomes are composed into an extensive resource and made freely accessible via mitophos.de.

**Results**

**Comprehensive mitochondria proteome coverage across various mouse tissues**

To advance our understanding of tissue-specific functional specializations of mitochondria at the protein level, we set out to characterize proteomes of mitochondria collected from various mouse tissues by LC-MS/MS analysis. To this end, we first isolated mitochondria from seven tissues—brain, BAT, heart, kidney, liver, SKM, spleen—from six 18–21-wk-old C57BL/6N mice (three females and three males). Mouse tissues were homogenized with a Dounce homogenizer, and crude mitochondria were isolated via differential centrifugation and subsequently purified on a Percoll density gradient to obtain ultra-pure mitochondria isolates (Kühlf et al, 2017). Importantly, this procedure was shown to efficiently exclude contaminations from other cellular compartments (Wieckowski et al, 2009). It was possible that some of the mitochondria would lose their integrity during our multistep enrichment procedure. However, previous work using similar procedures have specifically investigated this possibility and have concluded that this is not a substantial issue (see, for example, [Becker et al, 2019] and Fig 3 from Lu et al [2022]). Furthermore, the repeated removal of contaminating material (released proteins, cytosol, cell debris, other organelles) during differential centrifugation and the concentration of largely intact mitochondria in the Percoll gradient makes a strong contamination of randomly phosphorylated/dephosphorylated mitochondria unlikely.

Proteomes of these ultra-pure mitochondrial samples were acquired by a state-of-the-art proteomics workflow (Fig 1A), allowing the robust identification and quantification (coefficient of variation [CV] < 20%) of proteins that covered a dynamic range of more than five orders of magnitude (Supplemental Data 1). In total, we identified over 7,000 proteins, of which, 1,266 were annotated as mitochondrial proteins using MitoCarta3.0 (Rath et al, 2021) and the IMPI (http://impi.mrc-mbu.cam.ac.uk/) database. This essentially covers 92% or 86% of the mitochondrial proteome by the measure of MitoCarta3.0 and the IMPI database, respectively (Fig 1B). In addition, 309 of the identified proteins were described as mitochondria-associated proteins in the IMPI database. For further analysis, we filtered for proteins identified in more than half of the biological replicates in at least one tissue, which resulted in 1,221 mitochondrial proteins. This still represents over 90% of MitoCarta3.0 and 82% of the IMPI database. Proteins of these mitochondrial samples were validated by mass spectrometry (MS)-based quantitative proteomics (Fig 1C), allowing the robust identification and quantification (coefficient of variation [CV] < 20%) of proteins that covered a dynamic range of more than five orders of magnitude (Supplemental Data 1). In total, we identified over 7,000 proteins, of which, 1,266 were annotated as mitochondrial proteins using MitoCarta3.0 (Rath et al, 2021) and the IMPI (http://impi.mrc-mbu.cam.ac.uk/) database. This essentially covers 92% or 86% of the mitochondrial proteome by the measure of MitoCarta3.0 and the IMPI database, respectively (Fig 1B).
databases and highlights the deep and reproducible mitochondrial proteome coverage of this study (Fig 1C). Interestingly, more than half of these mitochondrial proteins were identified across all tissues, whereas only 7% were exclusively detected in one specific tissue (Fig 1C), confirming previous reports on mitochondrial proteomes by us and others (Mootha et al., 2003; Forner et al., 2006; Johnson et al., 2007; Calvo & Mootha, 2010). Of these, over 66% (60 proteins) were both reproducibly identified and not in the lowest 20% of ranked abundances (Supplemental Data 2), making them clear candidates for tissue-specific mitochondrial proteins. A similar proportion of the mitochondrial proteome was also exclusive to two or more tissues by the same criterion.

Mitochondrial enrichment efficiency can be determined by the proportion of summed signal intensity for mitochondrial proteins in relation to all identified proteins in measured samples (Williams et al., 2018). Applying this strategy, we determined the proportions to be very high (>90%) for liver, kidney, SKM, heart, and BAT, but lower for brain (63%) and spleen (47%) (Fig 2A). To further evaluate the mitochondrial purity, we performed a western blot analysis from sample aliquots of one mouse taken at various stages during the

Figure 1. Mitochondrial proteome and phosphoproteome preparation. (A) Workflow of tissue preparations for mitochondrial proteome and phosphoproteome enrichment, and LC-MS/MS analysis (n = 6). Tissues were first homogenized (* for skeletal muscle, see Materials and Methods section), crude mitochondria were isolated and ultrapure mitochondria were obtained using a Percoll gradient. Proteins were digested and prepared for phosphoproteome analysis via TiO2 enrichment or subjected to LC-MS/MS. (B) Distinct protein identification across biological replicates (n = 6). Annotation of proteins as mitochondrial is based on MitoCarta3.0 and the IMPI database. (C) Mitochondrial protein numbers after filtering for mitochondrial proteins identified in at least 50% of biological replicates (n = 6) in at least one tissue. Skeletal muscle (SKM), brown adipose tissue (BAT).

Source data are available for this figure.
Figure 2. Mitochondria enriched samples show tissue-specific clustering.

(A) Mitochondrial (red) and not mitochondrial (blue) proteins identified, based on the MitoCarta3.0 and IMPI database, are ranked by their intensity for each individual tissue. Histograms on the top and right display the distribution of mitochondrial (red) and not mitochondrial (blue) proteins along the rank and the intensity axes, respectively. The percentage of all identified mitochondrial (red) and not mitochondrial (blue) proteins and their summed intensities are displayed in bar graphs.

(B) Heatmap showing Pearson correlation coefficient for biological replicates (n = 6) for mitochondrial proteins of all mitochondrial enriched samples.

(C) Principal component analysis of mitochondrial proteins of all acquired biological replicates (n = 6). Skeletal muscle (SKM), brown adipose tissue (BAT).

Source data are available for this figure.
sample preparation process (Fig S1A). This analysis is consistent with the LC-MSMS measurements and supports the effective enrichment of mitochondria within the ultrapure mitochondria fraction. The absence of the cytosolic markers tubulin or actin (for brain) in both crude and ultrapure mitochondria fractions from most tissues indicates successful reduction of cytosolic contamination (Fig S1B). However, we did detect a tubulin signal in the pure mitochondria fraction from brain tissue, which may be attributed to the known interaction between mitochondria and microtubules, particularly during processes such as axonal transport (Lewis et al, 2018). Whereas potential contamination by other cellular compartments in brain and spleen tissue cannot be completely ruled out, these trends are consistent with the established literature (Williams et al, 2018; Fecher et al, 2019; McLaughlin et al, 2020; Roberts et al, 2021) and can likely be explained by the high tissue heterogeneity of brain (Fecher et al, 2019; Menacho & Prigione, 2020) and spleen. Indeed, spleen tissue consists of various types of immune cells (Lewis et al, 2019), which might impede high purity enrichment of its organelles. For the brain, contaminations by synaptosomes, which themselves contain neuronal mitochondria, have frequently been observed (Mootha et al, 2003). Yet, correlation of both mitochondrial and all identified proteins between biological replicates yielded Pearson correlation coefficients higher than 95% in all tissues (Figs 2B and S2A). Principle component analysis further shows that biological replicates cluster together and underlines functional similarities between tissues such as heart and SKM and tissue-related diversity of mitochondria proteomes (Figs 2C and S2B).

Mitochondrial proteome composition reveals tissue-specific functions

To gain further insights into tissue-specific functional differences in mitochondria, we investigated differences in the composition of mitochondrial proteomes across tissues. First, we focused on the oxidative phosphorylation system, which is essential for production of the energy-rich metabolite ATP and other processes like free radical generation and apoptosis (Hüttemann et al, 2007). We found that proteins of the electron transport chain (Complex I–Complex IV) and ATP-synthase (Complex V) displayed high abundances in heart and SKM tissues, supporting the physiological requirement of high ATP levels in both muscular tissues to sustain processes like muscle contraction (Ferreira et al, 2010; Ventura-Clapier et al, 2011) (Fig 3A). Conversely, levels of Complex V proteins were substantially lower in mitochondria from BAT compared with all other measured tissues (Fig 3A), which agrees with the specialized function of BAT in non-shivering thermogenesis (Jastroch et al, 2010; Kajimura & Saito, 2014; Oelkrug et al, 2015). This was further supported by the high abundance of the uncoupling protein 1 in our proteome measurements, the key mediator of the heat-generating proton leak in the mitochondria of BAT (Fig 3B).

In contrast to the well-described uncoupling protein 1, several other members of the SLC25 family remain uncharacterized (Ruprecht & Kunji, 2020). In our dataset, we identified a total of 47 members of the SLC25 family (Kunji et al, 2020). We found that levels of SLC25A23 and SLC25A25, two ATP-Mg\(^{2+}\)/carrier paralogues, (del Arco & Satrustegui, 2004; Fiermonte et al, 2004), displayed high abundance in the mitochondria of brain tissue (Fig 3C and D). Interestingly, knockout of SLC25A23 was shown to increase neuronal vulnerability (Rueda et al, 2015), which corroborates our observation and suggests an important role of SLC25A2 in this tissue type. We also quantified a third paralogue, SLC25A24, which showed a higher abundance in spleen tissue. Notably, the spleen harbors a large pool of B-cells and reduced SLC25A24 levels were previously linked to B-cell malignancies (Sandhu et al, 2013) (Fig 3E). Although these three ATP-Mg\(^{2+}\)/P carriers were not detected in the mitochondria from heart tissue, we identified another class of ATP carriers, including SLC25A4 or SLC25A31, which both showed increased abundance in the heart compared with all other measured tissues. Such differences between mitochondrial proteome compositions are easily retrieved from our dataset, facilitating a better understanding of mitochondrial plasticity across tissues.

Tissue specificity of mitochondria-associated proteins

The identification of key proteins mediating the crosstalk of mitochondria with their cellular environment is crucial to better understand their tissue-specific regulation and this concept has already attracted considerable interest in recent years (Montes de Oca Balderas, 2021). Although the characterization of local proteomes of organelar contact sites usually requires special centrifugation-based isolation methods, we anticipated that a considerable fraction of mitochondria-associated proteins would also enrich along with the mitochondria in our samples. We first performed an annotation term enrichment analysis of all identified proteins. Whereas we observed significant enrichment of several mitochondria-related terms in all tissues as expected, the non-mitochondrial protein pool was largely enriched for terms related to the tissue of origin (Fig S3, SourceDataForFigure4). For instance, terms like “positive regulation of B cell activation” in spleen or “positive regulation of synapse assembly” in brain tissue suggest tissue-specific functions. The enrichment of G-protein associated terms marks an exception, as these were enriched across all tested tissues.

Next, we performed a network analysis to evaluate the nature and quality of co-enriched proteins, more specifically whether non-mitochondrial proteins identified in these samples are associating proteins with functional roles or biological contaminations that are likely tissue-specific and highly abundant. This analysis revealed several clusters of known mitochondrial complexes such as the TIM23 complex or processes like the ubiquinone biosynthetic process, whose members were robustly identified in all tissues, and several clusters including both mitochondrial and non-mitochondrial proteins (Fig 4). For instance, one cluster consisted of G proteins, some of which were identified throughout all tissues and shown to localize to mitochondria (e.g., GNAI2) (Benincà et al, 2014), although not annotated as mitochondrial proteins by the IMPI or MitoCarta3.0 database. Interestingly, like most of the proteins in the G protein cluster, many G-protein–coupled receptors were exclusively identified in brain tissue. A prominent member of these brain-specific G protein coupled receptors is CNR1, which was reported to localize to mitochondria where it plays an important role in the regulation of memory processes through the modulation of the mitochondrial energy metabolism (Hebert-Chatelain et al, 2016). These examples underscore the
Figure 3. Tissue-specific protein contribution of SLC25 proteins to the mitochondrial proteome.
(A) Representation of the oxidative phosphorylation system including from left to right the electron transport chain (Complex I [blue], Complex II [violet], Complex III [red], Complex IV [green]), and ATP synthase (Complex V [yellow]). Radar plots show the relative contribution of the corresponding complex across the analyzed tissues to the overall mitochondrial composition (see Materials and Methods section for detailed description). (B–D) Normalized intensity (median of all log2-transformed mitochondrial proteins of a sample was subtracted from all log2 protein intensities of that sample) of the uncoupling protein 1 (UCP1), SLC25A23, SLC25A24, and (E) SLC25A25 across all analyzed tissues (black dots indicate individual identifications). (B–D) Protein abundance differences in relation to the reference tissue (BAT in warm colors).

potential of our dataset in facilitating the identification of previously unrecognized mitochondrial and mitochondria-associated proteins through a hypothesis-free approach. Another cluster consisting of mitochondrial and non-mitochondrial proteins contains enzymes of the kynurenine pathway involved in the catabolism of tryptophan. Similar to the mitochondrial transaminases, Ccbl2 and Aadat, Ccbl1 was identified throughout all tissues (Han et al, 2010). Interestingly, Ccbl1 has a cytosolic and a mitochondrial isoform and has been previously shown to localize to the mitochondrial matrix (Malherbe et al, 1995).

Mitochondrial kinases and phosphatases show tissue specificity

Posttranslational modification of proteins, specifically phosphorylation, plays a crucial role in the orchestration of mitochondrial protein function (Niemi & Pagliarini, 2021). However, almost no mitochondrial kinases with mitochondrial targeting sequences have been consistently reported and most kinases shown to associate with mitochondria have been found on or interact with the outer membrane (Kotrasová et al, 2021). Given our deep mitochondrial proteomes, we investigated relative abundances of kinases and phosphatases, which are annotated as or suggested to associate with mitochondria. In total, we obtained a dataset of 618 phosphorylation sites on 335 mitochondrial proteins (Fig 5C, SourceDataForFigure5). Of these high-confidence sites, 16% have previously not been reported in mice according to the PhosphoSitePlus database (Hornbeck et al, 2012). Strikingly, in contrast to the mitochondrial proteomes, 32% of the phosphosites identified on mitochondrial proteins were exclusive to one tissue, and only 9% were identified in all the tissues measured (Fig 5C). This might indicate that mitochondrial diversity is more strongly pronounced at the phosphorylation than the protein level.

To investigate whether there exists a correlation between the abundances of kinases and their documented phosphorylation sites across various tissues. The aforementioned PDK kinases were previously reported to regulate the phosphorylation of the phosphorylation sites S293, S300, and S232 on PDHA1 (Korotchkinsa & Patel, 2001). These authors further showed that PDK1 is predominantly responsible for the phosphorylation of S232 on PDHA1. In our data, there was no clear correlation between the kinase and its respective phosphorylation site (Fig 5C). The absence of a definitive link between PDK1 and S232 phosphorylation abundance in our dataset suggests the need to consider additional factors like signaling events or phosphatase activity across different tissues to comprehensively evaluate the phosphorylation status of distinct phosphosites. For instance, the phosphatases PDP1 and PDP2 measured here by proteomics both have the capability to dephosphorylate S293, S300, and S232 on PDHA1, albeit with varying efficiencies (Karpova et al, 2003). Another example for a known phosphatase—substrate relation exists between the phosphatase Pptc7 and the phosphorylation of T33 on Timm50 (Niemi et al, 2019). Although excluded by our stringent filtering criteria that require the presence of a phosphorylation site in five of six replicates, it is interesting to note that we observe low Pptc7 levels in liver, which is the only tissue where we detect Timm50 T33 phosphorylation, congruent with the phosphatase—substrate relationship.

Mitochondrial phosphoproteomes exhibit extensive intra-mitochondrial phosphorylation

To understand the distribution of mitochondrial phosphoproteins across mitochondrial compartments, we examined the sub-mitochondrial localization based on the curated MitoCarta3.0...
Mitochondria are typically divided into four main compartments, that is, mitochondrial outer membrane (OMM), intermembrane space, inner mitochondrial membrane (IMM), and matrix, although the complex organization of the IMM possibly may define additional compartments (Colina-Tenorio et al, 2020). In line with the high proportion of shared mitochondrial

Figure 4. Proteome of mitochondria enriched samples displays tissue-specific complexes. Cytoscape network analysis of reproducibly (>50% identification rate in at least one tissue, six replicates were analyzed [n = 6]) identified proteins of all mitochondria enriched samples. The main network depicts mitochondrial (orange) and non-mitochondrial (blue) proteins with at least one edge (String score >0.95). The size of individual nodes represents the number of identifications ranging from 3 (small circle) to 42 (big circle). Subnetworks display mitochondrial (orange) and non-mitochondrial (blue) proteins and tissues in which they were identified (dark green—brain; yellow—spleen; light green—liver; pink—kidney; ochre—SKM; blue—heart; orange—BAT). Skeletal muscle (SKM), brown adipose tissue (BAT).

Source data are available for this figure.
proteomes across all tissues (Fig 1C), the overall localization of proteins was not different between tissues and closely resembled the distribution of all annotated mitochondrial proteins in the database (Fig 6A). However, when performing the same analysis using the phosphorylated mitochondrial proteins, we observed a significant shift towards a localization to the OMM in all tissues (adj. \( P \)-values < 1.6 × 10^{-9}) (Figs 6A and S5). Surprisingly, our data also showed that depending on the tissue type, more than...
60% of phosphorylated mitochondrial proteins had an intra-mitochondrial annotation—IMM, intermembrane space or matrix localization—(Figs 6A and S5). Interestingly, 23–50% of OMM, but only 3–21% of intra-mitochondrial proteins was phosphorylated. Here, especially brain (8%) and spleen (4%) tissues showed low intramitochondrial phosphorylation rates. Given their unique evolutionary origins, we were interested in phosphorylation sites on proteins encoded by mitochondrial DNA, but failed to detect any (Table S2).

We further investigated the localization of specific kinase-substrate associations (KSA) across sub-mitochondrial localizations using NetworKin3.0 (Horn et al, 2014). Prominently, more than 40% of the predicted KSA in the IMM were linked to the PKC kinase family (Fig 6B). Studies have already reported the localization of PKC kinase family members to mitochondria, and an increased phosphorylation of the IMM protein COX IV after PKCe activation (Majumder et al, 2000; Baines et al, 2002; Ping et al, 2002; Jaburek et al, 2006; Ogbi & Johnson, 2006). Moreover, the MAPK group appeared to act on proteins localized to the OMM and matrix, whereas the PDHK family (including PDHK1–4 and Bckdk) was specifically associated with the matrix proteins (Fig 6B). The members of the latter kinase family are known to localize to the mitochondria matrix (Hitosugi et al, 2011), further supporting the validity of identified KSA. However, molecular studies are needed to investigate such KSA, whether phosphorylation of intra-mitochondrial proteins occurs in situ or outside mitochondria before being imported into mitochondria, how and which kinases/phosphatases translocate to or into mitochondria, and whether these phosphorylation events are functionally relevant.

**Mitochondrial phosphoproteome reveals tissue-specific modulation of fusion and fission events**

The tissue-specific phosphorylation of mitochondrial proteins suggested functional differences in mitochondria and prompted us to investigate the influence of phosphorylation on mitochondrial dynamics. We focused on proteins involved in mitochondrial fusion and fission, two important counteracting events involved in
organellar distribution, size balancing, and maintenance of a healthy mitochondrial network (Silva Ramos et al, 2019; Liu et al, 2020). Especially proteins involved in the fission process are regulated by a range of protein modifications, including phosphorylation (Cribbs & Strack, 2007; Taguchi et al, 2007; van der Bliek et al, 2013).

Throughout all tissues, MIGA1 (FAM73A) and MIGA2 (FAM73B), two homologues regulating mitochondrial fission by functioning downstream of the mitofusins, showed different abundances (Fig S6). This is especially interesting because MIGA1 and MIGA2 can form hetero and homodimers, highlighting a different regulation of fusion in different tissues (Zhang et al, 2016). Moreover, we identified multiple phosphorylation site clusters on MIGA2, whereas none were identified on MIGA1 (Fig S6). Intriguingly, similar phosphorylation clusters were identified on Miga in Drosophila melanogaster (Xu et al, 2020). Interestingly, two phosphorylation sites on Miga, S246 and S249, were reported to be essential for Vap33 interaction and the establishment of endoplasmic reticulum-mitochondria contact site (ERMCS), suggesting that phosphorylation on MIGA2 has similar functions in mammals (Xu et al, 2020).

Moreover, we observed that GTPase dynamin-related protein 1 (DRP1), a crucial player initiating mitochondrial fission (Bleazard et al, 1999; Cereghetti et al, 2008), displayed higher abundance in brain compared with other tissues (Fig 7A). This observation supports the importance of mitochondrial fission in neurons, where mitochondria switch to a fragmented morphology to enter and travel through axons (Lewis et al, 2018). We cannot exclude that the signal for DRP1 could be related to the lower mitochondrial purity in brain tissue. However, increased levels of serine 622 phosphorylation on DRP1, a site that has been shown to regulate DRP1 translocation to mitochondria (Cribbs & Strack, 2007; Taguchi et al, 2007; Cereghetti et al, 2008), suggests that DRP1 is actively localized to mitochondria, likely to regulate frequent fission events in the brain tissue (Fig 7B). Furthermore, we detected Erk2, a kinase capable of phosphorylating DRP1 at S622 (Kashatus et al, 2015), in both brain and spleen tissues (Fig S7A). Intriguingly, we also observed phosphorylation of Erk2 at residues T183 and Y185, sites known to be associated with the mitochondrial localization of Erk2 (Fig S7B) (Helfenberger et al, 2018).

In mammals, four DRP1 receptor proteins, that are all integral membrane proteins of the OMM, have been reported: mitochondrial fission protein 1 (FIS1), mitochondrial fission factor (MFF), and mitochondrial dynamics protein MiD49 (MIEF1) and MiD51 (MIEF2) (Losón et al, 2013). MFF and FIS1, the fission-promoting receptors, were robustly quantified in all tissues and displayed higher abundances in brain and spleen compared with other tissues (Fig 7A). However, MIEF1/2, which counteract DRP1-mediated fission (Dikov & Reichert, 2011; Liu et al, 2013), were generally too low to be robustly quantified in the measured tissues. In addition, we detected higher levels of MFF phosphorylation at the serine 129 and 146 residues in brain tissue compared with other tissues in which they were detected (Fig 7C and D). Both sites are essential for the recruitment of DRP1 and initiation of fission (Ducommun et al, 2015; Toyama et al, 2016; Lewis et al, 2018).

Elevated DRP1 levels have been shown to increase reactive oxygen species levels (Watanabe et al, 2014). Intriguingly, we identified oxidation resistance 1 (OXR1), exclusively in the mitochondria of brain tissue, where it plays an important role in the protection of neuronal cells from oxidative stress (Volkert & Crowley, 2020). This likely indicates a protective function of OXR1 in brain tissue as a response to the prevalent fragmented organellar morphology induced by DRP1. In addition, we found OXR1 to be hyper-phosphorylated and three out of 12 high-confidence sites were novel (Fig 7E). Tissue specificity and the lack of functional annotation of phosphorylation sites that are identified in our study make OXR1 an exciting candidate to be investigated in the future.

**Web application makes Mouse Mitochondria Atlas data readily accessible**

As indicated by the above examples, this study presents a rich resource to explore the mitochondrial proteomes and phosphoproteomes across mouse tissues. Preceding examples show the potential of this resource for investigation of tissue-specific mitochondrial regulations on the proteome and phosphoproteome levels, ultimately permitting the generation and analysis of new hypotheses. However, use of such resources largely depends on the ease of data access for exploration.

To this end, we created a web application mitophos.de offering the end user an interface to easily explore datasets, including MitoCarta3.0 networks, abundance comparisons across tissues, and sequence analysis (Fig 8A). As an example, Fig 8 illustrates the MICOS complex, which has a central role in mitochondria (Khosravi & Harner, 2020). In the network view, one can see (I) members of the complex and the phosphorylation sites identified on these proteins and (II) in which tissues and how reproducibly they are identified in our dataset (Fig 8B). Moreover, the user can inspect individual abundance distributions of all identified proteins/STY sites across all measured tissues. For instance, MINOS1, a core component of the MICOS complex (Bohnert et al, 2015), displayed high abundance in mitochondria isolated from heart, SKM, and BAT tissues (Fig 8C). Moreover, in the sequence view the AlphaMap tool (Voytik et al, 2022) is integrated to map all identified peptides, including phosphorylated peptides, onto their respective protein sequence along with structural information such as topological domains and transmembrane regions (Fig 8D) and to visualize phosphorylation sites in their 3D structures (unpublished data) as predicted by AlphaFold (Jumper et al, 2021).

Together, this data-rich and comprehensive tool is an entry point to investigate the herein presented resource and will assist in future efforts to functionally characterize mitochondrial proteins and their respective phosphorylation sites.

**Discussion**

Here, we present a tissue-specific atlas of mouse mitochondrial proteomes and phosphoproteomes, an in-depth resource towards a better understanding of the composition and function of this vital organelle in a tissue-specific manner. Previous MS-based studies combining mitochondrial phosphoproteome and proteome measurements typically focused on a single or few tissues and were generally shallower than our study. In addition,
differences in study designs such as the use of various organisms or mitochondria/phosphopeptide enrichment protocols, analysis pipelines, and mitochondrial protein annotation databases, complicate the integration of such datasets to understand tissue specificity. We now globally and precisely quantified different protein expression and phosphorylation patterns at subcellular level across seven mouse tissues, providing a detailed view on mitochondrial diversity. The breadth and depth of coverage achieved...
Figure 8. Web application readily enables easy data access. 

(A) Scheme of Mouse Mitochondria Atlas application features (mitophos.de). (B) MICOS complex view based on MitoCarta3.0 annotation. Large and small nodes represent proteins and class 1 STY sites, respectively. Edges represent on String interaction scores >0.4 and color of nodes indicate the number of tissues in which proteins/STY sites were identified. 

(C) Normalized intensities (median of all log2 transformed mitochondrial proteins of a sample was subtracted from all log2 protein intensities of that sample) for Minos1 across all analyzed tissues (black dots indicate individual identifications, box and error bar). (D) Sequence plot of Micos1 shows structural information (based on UniProt annotations) and protein coverage based on identified peptides. All identified STY sites are marked with a star. Data in this figure are based on the analysis of six replicates (n = 6). Skeletal muscle (SKM), brown adipose tissue (BAT).
by the integration of tissue-specific mitochondrial proteomes and phosphoproteomes provide unbiased insights into the mitochondrial composition and function. This allows the generation and assessment of novel hypotheses related to mitochondrial biology, which cannot be generated with focused studies alone. Integrated mitochondrial proteomes and phosphoproteomes that are diverse between tissues can readily be explored at mitophos.de.

Our data revealed that the functional diversity of mitochondria is defined by protein abundance rather than compositional differences because more than half of the mitochondrial proteome was shared by all analyzed tissues and 90% by at least two tissues. For instance, the electron transport chain is an integral part of the mitochondrial composition and its components are found across all tissues; however, their abundance shows substantial differences to meet tissue-specific energy demands. Thus, dysregulation of individual proteins can strongly affect mitochondria in one tissue, leading to severe diseases, whereas mitochondria in a different tissue remain largely unaffected. Moreover, we found that 7% of the proteome displays tissue specificity, further contributing to our understanding of tissue-specific effects of mitochondrial protein dysregulation (Russell et al, 2020). This is an important concept, which can now be studied in an unbiased manner using our resource data, providing opportunities for development of targeted treatments for mitochondrial diseases. For instance, members of the SLC25 family are linked to metabolic diseases in distinct tissues (Palmieri & Monne, 2016) and various cancers (Rochette et al, 2020). However, biological functions of a large repertoire of mitochondrial SLCs are still unknown. For example, inactivation of SLC25A25 was assessed in mouse SKM tissue where it caused a reduced metabolic efficiency (Anunciado-Koza et al, 2011). Given its high abundance in the mitochondria of brain tissue and in glioma cells (Traba et al, 2012), it will also be interesting to investigate its role in this tissue, particularly whether SLC25A25 deficiency influences neuronal fitness.

Mitochondria are essential cellular entities that are involved in a wide variety of cellular processes (McBride et al, 2006) through dynamic interaction and constant communication with other organelles such as the (ER), nucleus, and peroxisomes via membrane contact sites (Śhai et al, 2018; Desai et al, 2020; Perrone et al, 2020) or protein complexes, such as the ribosome (Lashkevich & Dmitriev, 2021). We suggest that non-mitochondrial proteins identified in the samples might present signatures that could convey important biological information regarding mitochondria-associated structures. Given the high level of mitochondrial enrichment combined with highly reproducible LC-MS/MS measurements, such protein signatures are unlikely to be solely based on unspecified enrichment of abundant proteins. For example, the proteasome is robustly identified in most of the mitochondria-enriched samples, which is in line with its involvement in the degradation of misfolded mitochondrial proteins (Basch et al, 2020; Kodron et al, 2021). We also observed multiple G-proteins that were previously shown to interact with mitochondria, demonstrating the relevance of mitochondria in cellular signaling processes. Furthermore, our data identified non-mitochondrial ribosomal proteins in all tissues, which could be explained by the local translation of nuclear-encoded mitochondrial mRNAs (Lashkevich & Dmitriev, 2021). It was recently shown that RNA-bearing late endosomes associate with mitochondria and ribosomes forming hotspots of local protein synthesis in axons (Cioni et al, 2019) and that mitochondria fuel such local translations in neurons, enabling synaptic plasticity (Rangaraju et al, 2019).

There is mounting evidence that phosphorylation of mitochondrial proteins fulfills important functions to maintain cellular health as exemplified by the fission process in this study. Deregulation of mitochondrial protein phosphorylation can lead to diseases such as cancer, diabetes, heart and neurological disorders. It was recently reported that 91% of mitochondrial proteins on MitoCarta3.0 have at least one phosphorylation site reported on the PhosphoSitePlus database (Niemi & Pagliarini, 2021). However, this analysis also includes proteins that do not always localize to mitochondria (Ben-Menachem et al, 2011) and phosphorylation sites that can specifically be captured upon perturbations and stimuli. Our study revealed that around one third of the mitochondria-localized proteins were phosphorylated in tissues at steady state. This suggests that the mitochondrial proteome and phosphoproteome compositions are dynamically modulated in response to environmental changes. Furthermore, we identified 57 kinases and 11 phosphatases that either are localized to mitochondria or associate with mitochondria, providing a global view on important modulators of mitochondrial protein phosphorylation. Mapping tissue-specific mitochondrial kinases and phosphatases is an important step towards understanding their role in the regulation of the mitochondrial phosphoproteome in different tissues and hence developing therapeutics for mitochondrial diseases. For example, a mitochondrial phosphatase, phosphoglycerate mutase family member 5 (PGAM5), has recently emerged as an important regulator of mitochondrial homeostasis. Deletion of PGAM5 has been shown to result in Parkinson’s-like movement disorder in mice (Lu et al, 2014) and T cell dysfunction in primary cells (Panda et al, 2016). Although its diverse roles largely remain to be uncovered (Liang et al, 2021), a novel PGAM5 inhibitor was recently suggested as a potential therapeutic for brain ischemic stroke (Gao et al, 2021). We observed that PGAM5 displays elevated levels in the mitochondria isolated from brain, SKM, and spleen, possibly explaining the tissue-specific phenotypes induced by its absence and where in the body molecules targeting its phosphatase activity would exert their effects.

The data in this study will contribute to our understanding of the tissue-specific composition and function of mitochondria and serve as a gateway for investigation of specific questions related to mitochondrial biology. Future biochemical and more focused investigations are needed to validate our findings and test the hypotheses arising from our study. For instance, it is pivotal to experimentally validate KSA of previously unknown phosphorylations on mitochondrial proteins and their functional implications in the cell. In addition, the impact of those phosphorylations on the localization of target proteins and, more specifically, the question of whether mitochondrial proteins are phosphorylated before or after entering the mitochondria remain to be investigated. Accessibility, for instance, of previously undescribed phosphorylation sites on mitochondrial proteins can be assessed using advanced structural tools (Jumper et al, 2021) to determine if they are likely targeted by a mitochondrial kinase or a
cytoplasmic kinase before being imported (Rosenberger et al, 2021). Moreover, future developments towards better enrichment strategies for the isolation of mitochondria from different tissues and advances in the MS technology will aid to further improve the depth and quality of the mitochondrial proteomes and phosphoproteome.

**Materials and Methods**

**Experimental model and subject details**

Six C57BL/6N mice (three male, three female) were housed in a 12-h light/dark cycle in standard ventilated cages under specific pathogen-free conditions with constant temperature (21°C) and humidity (50–60%) and fed ad libitum with a standard mouse diet. At the age of 18–21 wk, mice were euthanized by cervical dislocation. The study was approved by the by the Landesamt für Natur, Umwelt und Verbraucherschutz Nordrhein-Westfalen, Germany, and performed in accordance with European law.

**Tissue preparation and isolation of ultrapure mitochondria**

Mice were euthanized by cervical dislocation and the seven tissues—heart, SKM, BAT, spleen, kidney, liver, and brain—were rapidly removed. Heart, spleen, and kidney were homogenized in mitochondrial isolation buffer containing 320 mM sucrose, 1 mM EDTA, and 10 mM Tris–HCl, pH 7.4, supplemented with EDTA-free complete protease inhibitor cocktail and PhosSTOP tablets (Roche). For isolation of mitochondria from BAT, liver, and brain, the mitochondrial isolation buffer was additionally supplemented with 0.2% BSA (Sigma–Aldrich). Subsequently, crude mitochondria were isolated from the homogenates by two rounds of differential centrifugation (see Fig 1A). Isolation of crude mitochondria from SKM was performed as previously described (Frezza et al, 2007). Crude mitochondrial pellets from all tissues were further purified on a Percoll density gradient as described recently (Kühl et al, 2017). Briefly, mitochondrial pellets were washed once in 1xM buffer (220 mM mannitol, 70 mM sucrose, 5 mM HEPES pH 7.4, 1 mM EGTA pH 7.4; pH was adjusted with potassium hydroxide; supplemented with EDTA-free complete protease inhibitor cocktail and PhosSTOP tablets [Roche]); and subsequently purified on a Percoll (GE healthcare) density gradient of 12%–19%–40% via centrifugation in a SW41 rotor at 43,000 g for 30 min in a Beckman Coulter Optima L-100 XP ultracentrifuge using 14 × 89 mm Ultra-Clear Centrifuge Tubes (Beckman Instruments Inc.). Ultrapure mitochondria were harvested at the interphase between 19% and 40% and washed three times with 1xM buffer. Dry mitochondrial pellets were snap-frozen in liquid nitrogen and stored at −80°C until further use.

**Western blot and antisera**

The protein concentration of protein samples was determined using the RCDC assay (Bio–Rad) and BSA as a standard. For standard gel electrophoresis, protein samples (40–60 μg/lane) were mixed with 2x NuPAGE LDS sample buffer supplemented with 200 mM DTT, heated for 10 min, and resolved using commercially available 4–12% NuPAGE Bis–Tris gels and MOPS buffer (Invitrogen) including protein standards (Spectra Multicolor Broad Range; Thermo Fisher Scientific). Proteins were transferred on polyvinylidene difluoride membranes (GE Healthcare) using wet tank blotting (25 mM Tris, 192 mM glycin, and 20% ethanol) at 4°C overnight at 80 mA. Membranes were incubated for 15 min in Ponceau 5 solution (Sigma–Aldrich) or Coomassie Blue, and blocked in 5% milk–1x TBS–0.1% Tween 20 (TBST) for at least 1 h at RT. Membranes were subsequently incubated with primary antibodies diluted in 3% milk–TBST overnight at 4°C, washed in TBST and incubated with HRP-conjugated secondary antibodies for 2 h at RT. After washing with TBST, immunodetection was performed by enhanced chemiluminescence (Immun-Star HRP Luminol/Enhancer from Bio–Rad) using the ChemiDoc Imaging System (Bio–Rad) and ImageLab software (Bio–Rad). The following antibodies were used to verify purity of different fractions: actin (ab8224; Abcam), histone 3 (Sigma–Aldrich), GADPH (GTX627408; GeneTex), PDI (GTX30716; GeneTex), TFAM (ab131607; Abcam), tubulin (clone 11H10, 2125; Cell Signaling), VDAC1 (clone N1528/23, MABN504; Millipore).

**Mitochondrial (phospho)proteome sample preparation**

Frozen ultrapure mitochondria pellets were resuspended in lysis buffer (4% SDC, 100 mM Tris/HCl, pH 8.5), boiled for 5 min at 95°C, and sonicated in 30 s intervals for 15 min (Bioruptor). Protein concentration was estimated via Tryptophan assay (Kulak et al, 2014) and was adjusted with lysis buffer to a total volume of 270 μl containing 400 μg of protein for brain, SKM, liver, heart, kidney, and 140 μg of protein for BAT samples. Proteins were reduced and alkylated by adding 30 μl of 10x reduction/alkylation solution 100 mM Tris (2-carboxyethyl)phosphine hydrochloride and 600 mM 2-chloroacetamide, followed by 5 min incubation at 45°C. Subsequently, 1:100 trypsin and LysC were added for overnight protein digestion at 37°C. For proteome analysis, 10 μl (brain, SKM, liver, heart, kidney) and 20 μl (BAT, spleen) aliquots were taken and loaded on SDB-RPS StageTips. Peptides were washed with 200 μl wash buffer (0.2% TFA/2% ACN [vol/vol]) and then eluted with SDB-RPS elution buffer (125% NH₄OH, 80% ACN [vol/vol]) and dried in a SpeedVac. Dried peptides were resuspended in A* buffer (2% ACN/0.1% TFA).

The remaining samples were processed following the EasyPhos protocol for phosphopeptide enrichment (Humphrey et al, 2018). In brief, samples were first mixed with isopropanol and EP buffer (48% TFA, 8 mM KH₂PO₄), followed by phosphopeptide enrichment with 5 mg TiO₂ beads per sample (GL Sciences). For this, samples were mixed with TiO₂ beads in loading buffer (6% TFA/80% ACN [vol/vol]) at a concentration of 1 mg/μl and incubated on a ThermoMixer (Eppendorf) for 5 min at 40°C by shaking at 1,200 rpm. Subsequently, beads were washed four times with 1 ml of wash buffer (5% TFA, 60% isopropanol [vol/vol]) and phosphopeptides were eluted from beads using 60 μl of elution buffer (40% ACN, 5% NH₄OH) and concentrated in a SpeedVac for 30 min at 45°C. Samples were immediately diluted with 100 μl of SDBRPS loading buffer (99% isopropanol, 1% TFA [vol/vol]) and loaded on SDB-RPS StageTips.
Thereafter, phosphopeptides were washed and eluted as described above and resuspended in 6 µl A*.

**LC-MS/MS**

For all measurements, peptides were loaded onto a 50-cm, in-house packed, reversed-phase column (75 µm inner diameter, 1. diameter, ReproSil-Pur C18-AQ 1.9 µm resin [Dr. Maisch GmbH]) and separated with and binary buffer system consisting of buffer A (0.1% formic acid [FA]) and buffer B (0.1% FA in 80% ACN). The column temperature was controlled by a homemade column oven and maintained at 60°C. For nanoflow liquid chromatography, an EASY-nLC 1200 system (Thermo Fisher Scientific), directly coupled online with a Q Exactive HF-X (Thermo Fisher Scientific) via a nano-electrospray source, was operated at a flow rate of 300 and 350 nL/min for mitochondrial proteome and phosphoproteome measurements, respectively.

For mitochondrial proteome measurements, 500 ng of peptides were loaded and separated using a gradient starting at 5% buffer B, increasing to 30% buffer B in 80 min, 60% buffer B in 4 min and 95% buffer B in 4 min. The MS was operated in DDA mode (Top12) with a full scan range of 300–1,650 m/z and a MS1 and MS2 resolution of 60,000 and 15,000, respectively. The automatic gain control was set to 3 × 106 and 1 × 105 for MS1 and MS2, whereas the maximum injection time was set to 20 and 60 ms, respectively. Precursor ion selection width was kept at 1.4 m/z and fragmentation was achieved by higher-energy collisional dissociation (NCE 27%). Dynamic exclusion was enabled and set to 20 s.

For mitochondrial phosphoproteome measurements, 5 µl as loaded and separated using a gradient starting at 3% buffer B, increasing to 19% buffer B in 40 min, 41% buffer B in 20 min, and 90% buffer B in 5 min. The MS was operated in DDA mode (Top10) with a full scan range of 300–1,600 and a MS1 and MS2 resolution of 60,000 and 15,000, respectively. The automatic gain control was set to 3 × 106 and 1 × 105 for MS1 and MS2, whereas the maximum injection time was set to 120 and 60 ms, respectively. Precursor ion selection width was kept at 1.6 m/z and fragmentation was achieved by higher energy collisional dissociation (NCE 27%). Dynamic exclusion was enabled and set to 30 s.

**Bioinformatics analysis**

Data analysis was performed using the python programing language using python (3.8.12) and the following packages: alphamap, matplotlib (3.5.0), mygene (3.2.2), numpy (1.19.2), pandas (1.1.3), pyteomics (4.3.3), requests (2.26.0), sciPy (1.7.2), seaborn (0.11.2), sklearn (0.0), upsetplot (0.6.0). All notebooks used for data analysis are available at GitHub (https://github.com/MannLabs). Identified proteins were filtered for at least three valid values in at least one tissue. Similarly, phosphorylation sites were filtered for at least five valid values in at least one tissue and a localization probability >75%. The UniProt API was used to map “ACC + ID” protein group identifiers provided by the MQ analysis to “ENSEMBL_ID,” “P_ENTREZGENEID” and “STRING_ID” for further analysis. “ENSEMBL_ID” and “P_ENTREZGENEID” identifier were used for mitochondrial protein annotation based on the IMPI (IMPl_2021_Q4, downloaded 27. October 2020) and Mitocarta3.0 (downloaded 1. January 2021) databases, respectively. Network analysis and visualization were performed with the StringApp (1.6.0) in Cytoscape (3.8.2). Kinase annotations are based on manual annotations (SourceDataForFigure6) and “pkinfam” (downloaded 9. April 2021). Networkin3.0 was used for KSA predictions, whereas the Networkin score was set to 1. Gene Ontology (GO) annotations for the Geno Ontology biological process term enrichment analysis were retrieved from UniProt (accessed 8. March 2021). Enrichment analysis was performed in Perseus (1.6.7.0) against the set of identified proteins in the corresponding tissue and the results were filtered for an intersection size >10. Missing values were only imputed for principle component analysis and heatmap analysis. For this, a Gaussian normal distribution with a width of 0.3 relative to the SD of measured values and a downshift of 1.8 SDs were used. For data normalization, intensity values were log2 transformed and then filtered for known proteins. The median value of these mitochondrial proteins was subtracted from all log2 transformed values. Significance testing for individual proteins and phosphopeptides as shown in Figs 3 and 7 was performed with the ordinary one-way ANOVA method or by two-sided t tests in GraphPad Prism (9.3.1) (SourceDataForFigure3, SourceDataForFigure7). Significance testing for differences in mitochondrial protein, phosphoprotein, and phosphosite localization was performed in RStudio (1.3.1093) (SourceDataForFigure6). OMM proportions were used for fitting a beta-regression model using the betareg R package with default settings (Cribari-Neto & Zeileis, 2010; Grun et al, 2012). P-values were estimated with the lrttest function of the lrttest package (Zeileis & Hothorn, 2015) and P-values were adjusted with the “fdr” method of the P.adjust function of the stats base package (Benjamini & Hochberg, 1995).

**Website tool**

The website tool is structured into four sections. The first three “Pathway view,” “Sequence view,” and “Tissue comparison” are for displaying data, whereas the fourth section provides explanations for each individual section. In the “Pathway view” and “Tissue comparison,” proteome and phosphoproteome data filtered for at least three and five identifications in at least one tissue,
respectively. Intensity values were normalized as described above and used for data representation in the “Tissue comparison” tab or z-scored across tissues and used for the “Pathway view tab.” Here, median z-score values of the six biological replicates per tissue are displayed in the data table. The polar plot represents the median z-score of all pathway/complex members of a given tissue. Network/complex annotations were retrieved from MitoCarta3.0 and protein interactions are based on STRING interaction scores. These STRING interaction scores were retrieved from STRING (17. November 2021) using the "STRING ID" and the STRING API. For the “Sequence view,” the "evidence.txt" of the MaxQuant output files was directly used as input to annotated sequences.

The python programming language was used for data processing and visualization for the dashboard. The following libraries were used for data processing: numpy (1.19.2), pandas (1.19.2), re, sys, os, and pyteomics (4.3.3). Several libraries from the HololVis family of tools were used for data visualization and creation of the dashboard, including panel (114.6), holoviews (114.6), bokeh (2.2.2), plotly (4.12.0), and param (1.10.0). Network visualization was achieved with the NetworkX package (Hagberg et al, 2008). The Alphamap tool (Voytik et al, 2022) was integrated to display linear protein sequence annotations and to visualize 3D protein structures.

Data Availability

Datasets generated in this study have been deposited at ProteomeXchange and are publicly available (Identifier: PXD030062). All original code has been deposited on GitHub (https://github.com/MannLabs) and is available as of the date of publication. Any additional information required to reanalyze the data reported in this article is available from the lead contact upon request.

Supplementary Information

Supplementary Information is available at https://doi.org/10.26508/lsa.202302147.

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Conflict of Interest Statement

The authors declare that they have no conflict of interest.

References


Author Contributions

FM Hansen: conceptualization, software, formal analysis, investigation, visualization, methodology, and writing—original draft, review, and editing.

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I Bludau: data curation, software, methodology, and writing—review and editing.

N-G Larsson: conceptualization, resources, supervision, funding acquisition, project administration, and writing—review and editing.

I Kühl: conceptualization, investigation, methodology, and writing—review and editing.

M Mann: conceptualization, supervision, funding acquisition, project administration, and writing—original draft, review, and editing.

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