Proximity labelling reveals the compartmental proteome of murine sensory neurons

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Abstract

Understanding the molecular architecture of peripheral sensory neurons is critical as we pursue novel drug targets against pain and neuropathy. Sensory neurons in the dorsal root ganglion (DRG) show extensive compartmentalization, thus understanding each compartment – from the peripheral to central terminals – is key to this effort. To systematically profile this spatial complexity, we generated a TurboID[fl/fl] transgenic mouse line (ROSA26[em1(TurboID)Bros]), enabling targeted proximity labelling and deep proteomic profiling of DRG neuron compartments via Advillin[Cre]. Our data reveal distinct proteomic signatures across neuronal compartments that reflect specialized neuronal functions. We provide proteomic insights into previously inaccessible nerve terminals both in the periphery (innervating the skin) and in the spinal cord. Further, using a DRG explant model of chemotherapy-induced peripheral neuropathy (CIPN), we uncovered novel and discrete proteome changes, highlighting neuronal vulnerability. Together, our findings provide a unique proteome atlas of the sensory neuron proteome across anatomical domains and demonstrate the utility of proximity labelling proteomics for detecting compartment-specific molecular alterations in a disease model. This dataset serves as a valuable resource for mechanistic studies of sensory neuron function and pathology.

Introduction

Dorsal root ganglia (DRG) serve as critical hubs in the peripheral nervous system (PNS), housing the somata of sensory neurons that transmit information from the periphery to the central nervous system (CNS).[1] DRG comprise sensory neurons as well as other cells, including glial and immune cells that provide essential support for neuronal function and are implicated in PNS pathologies. Consequently, the interpretation of bulk proteomics data from DRG is inherently complex and requires downstream validation to discern cell type-specific changes.

Adding to this complexity, sensory neurons exhibit remarkable spatial compartmentalization. Understanding the spatial proteome of sensory neurons and its dynamics is critical for elucidating the molecular mechanisms underlying sensory processing, neuronal maintenance, and pathological conditions such as chronic pain. Somata are located far from their target organs and afferent synapses in the CNS, connected via axons that can extend up to more than a meter in length. Obtaining a comprehensive, cell-type- and compartment-specific proteomic profile of DRG neurons remains a significant technical challenge. Although techniques like fluorescence-activated cell sorting (FACS) can isolate specific cell types, they prioritize somata and/or nuclei. Isolating intact subcellular compartments, particularly the peripheral terminals/"free nerve endings", as well as the axons of sensory neurons remains a major hurdle.

In recent years, proximity-dependent labelling techniques combined with mass spectrometry (MS), such as TurboID-based biotinylation-dependent labelling and APEX, have emerged as powerful tools for mapping subcellular proteomes and interactomes in living cells and tissues.[2,3] TurboID is an engineered, promiscuous biotin ligase that enables rapid and efficient labelling of proteins within a 9–10 nm radius in living cells.[4] To date, most applications of TurboID and its predecessor BioID have involved fusion proteins with compartment-specific targeting sequences, enabling selective biotinylation of proximal interactomes for downstream affinity purification and identification via MS.[5–7]

Unlike TurboID fusion proteins, the use of soluble TurboID variants enables unbiased labelling of the entire accessible proteome within a cell, providing a comprehensive snapshot of its proteomic composition. Sunna et al. demonstrated that TurboID-mediated biotinylation covers approximately 59–65% of the proteome in neuroblastoma (N2A) and microglia (BV2) cell lines expressing a cytosolic TurboID variant.[8] By comparing the biotinylated proteome of these two cell lines, they revealed cell-type-specific proteome differences and captured proteome dynamics in response to an inflammatory challenge *in vitro*.

To explore cell-type-specific *in vivo* protein biotinylation and identification, Rayaprolu et al. developed an adenoviral and transgenic mouse model expressing TurboID in distinct

cell types, including neurons and astrocytes, across different brain regions, thereby uncovering region-specific proteomic differences in astrocytes.[9] In another *ex vivo* study, Hobson et al. employed APEX2 proximity labelling combined with MS to map the proteomic differences between somatodendritic and axonal compartments in acute brain slices.[10] They demonstrated how the compartmental proteome correlates with respective neuronal functions and identified ion channels with previously unrecognized axonal localization. This provided deeper insights into the molecular architecture of dopaminergic neurons.[10] These findings highlight the potential of proximity-dependent labelling approaches for mapping the proteome of spatially complex structures such as DRG neurons.

Here, we applied TurboID-based proximity labelling *in vivo* using a Cre-lox system. Paired to an Advillin[Cre] driver[11], this restricts TurboID expression to DRG sensory neurons. We were able to systematically profile the proteomes of distinct neuronal compartments, including somata (lumbar levels 3-5 DRG (L3-L5)), central synaptic terminals (terminals$_c$) in the dorsal lumbar spinal cord (SC), axons (sciatic nerve; SCN), and peripheral terminals (terminals$_p$) from paw skin, providing the first compartmentalized proteomic atlas of lumbar sensory neurons.

We then harnessed this technique to explore neuronal changes after oxaliplatin treatment in an *in vitro* explant model of chemotherapy-induced peripheral neuropathy (CIPN). CIPN is a common side effect of chemotherapy, with agents damaging sensory neurons and leading to a painful neuropathy often preventing adequate cancer therapy.[12] There are no preventative treatments available to date, thus a deeper understanding is urgently needed. We observed distinct dysregulation across neuronal transporters after oxaliplatin treatment, including some not detected with bulk proteomics. Several of these transporters represent promising candidates for follow-up translational studies.

Together, our findings reveal novel insights into the compartment-specific protein distributions that underlie key aspects of sensory neuron function. Our data also highlight the potential of TurboID-based proteomics to uncover subtle, cell type-specific changes in disease contexts. These datasets serve as valuable resources for future studies on neuronal specialization and pain-related proteomic alterations in the PNS.

Results

We profiled the compartmental proteome of murine sensory neurons using the well characterized Advillin[Cre] driver[11,13–15] crossed with the here generated TurboID[fl/fl] (Adv[Cre];TurboID[fl/fl]) mice (Figure 1). In these transgenic mice, TurboID is expressed in the cytoplasm of Advillin-positive cells, resulting in promiscuous biotinylation of intracellular proteins specifically within sensory neurons. Advillin-negative cells do not express the TurboID ligase and therefore do not label proteins (Fig 1A). We isolated tissue from the spinal cord (SC), dorsal root ganglia (DRG), sciatic nerve (SCN) and paw skin to cover the spatial complexity of DRG sensory neurons, isolating tissue from wildtype mice in parallel (Fig 1B). Biotinylated proteins were enriched after tissue lysis and prior to mass spectrometry, enhancing neuronal protein specificity in comparison to control tissue to account for endogenous biotinylation. Distinct labelling in each tissue of Adv[Cre];TurboID[fl/fl] mice was confirmed by immunohistochemistry (Fig 1C, SFig 1A) and validated via Western blotting (SFig 1B).

As expected, we see a clear enrichment of TurboID labelling in sensory neuron terminals in the dorsal root lamina, in neuronal cell bodies in the DRG, axons in the nerve, and nerve endings in the skin (Fig 1C). Using Western blotting, TurboID signal was readily detected in lysates from SC, DRG, and nerve tissues, but less prominent in skin lysates (SFig 1). The latter may be due to low protein quantity after lysis, as TurboID+ afferents are evident by immunohistochemistry (Fig 1C, "skin").

To generate a comprehensive and quantitative compartmental proteome atlas of DRG neurons, we employed data-independent acquisition combined with parallel accumulation–serial fragmentation (DIA-PASEF) mass spectrometry on Advillin[Cre] x TurboID (Adv[Cre];TurboID[fl/fl]) and control tissue (Advillin-negative mice, referred to as TurboID[fl/fl]) (Fig 2, SFig 2-5, STables 1-10). Both overlapping and unique proteins were detected across all 4 DRG compartments (Fig 2A). A comparison of Adv[Cre];TurboID[fl/fl] samples with TurboID[fl/fl] control shows more protein (SFig 2A), more protein groups (SFig 2B), and clustering by condition and compartment (SFig 2C-D). Adv[Cre];TurboID[fl/fl] samples cluster by compartment (Fig 2B), and show extensive ion channel coverage, including Trpv1 and Scn10a in the peripheral terminals (Fig 2C). These data highlight the sensitivity and depth of our profiling approach. A full expression matrix is available in STable 1.

Neuronal compartment proteome

In Adv[Cre];TurboID[fl/fl] samples proteins were considered enriched based on two criteria: A: proteins detected in Adv[Cre];TurboID[fl/fl] samples, but not in TurboID[fl/fl] samples, or B: proteins significantly upregulated in Adv[Cre];TurboID[fl/fl] samples compared to TurboID[fl/fl] ($\log_2$ fold change (LFC) > 1, FDR < 0.05, using a moderated T-Test through limma) (Fig 2D, STable 2). These upregulated proteins are referred to as differentially expressed

proteins (DEPs). The enriched proteins were then combined, and treated equally for all
downstream analyses.

We identified the highest number of enriched proteins in the somata (DRG), followed by
the central terminals (terminals$_c$, SC), axons (SCN), and finally the peripheral terminals
(terminals$_p$, skin), with overlapping and unique proteins (Fig 2E). For each compartment,
we see an enrichment of proteins in the Adv^{Cre};TurboID$^{fl/fl}$ sample over TurboID$^{fl/fl}$ (Fig 2F)
across the dynamic range (SFig 2E). In addition, there are few negatively correlated
proteins, likely reflecting endogenous biotinylation such as in the case of Acaca, Acacb,
Mccc1, and Pcca (Fig 2F).[16,17] Together, this results in a curated, compartmental
proteome for murine primary sensory neurons ranging from hundreds to thousands of
proteins per compartment.

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Figure 1. Adv^{Cre};TurboID$^{fl/fl}$ mice were used to target the sensory neuron proteome. A. Schema
overview for biotinylation of neuronal proteins. Adv^{Cre};TurboID$^{fl/fl}$ and TurboID$^{fl/fl}$ were provided
biotin-enriched water, which, in Adv^{Cre};TurboID$^{fl/fl}$ promote the biotinylation of all proteins
within Advillin-expressing (+) cells. Advillin-negative (-) cells only contain endogenously
biotinylated proteins. Tissues were lysed, proteins purified for biotin, prior to SP3 and
downstream LC-MS/MS via timsTOF HT in DIA-PASEF mode. B. Corresponding bulk proteomes

were isolated from wildtype (WT) mice for each corresponding tissue in A. C. Representative
immunohistochemistry from TurboID[fl/fl] (control) and Adv[Cre];TurboID[fl/fl] mice across tissues.
Cyan, DAPI. Magenta, Streptavidin-555 (biotin). Scale bar = 50 μm.
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Neuronal subtype enrichment

The use of the Advillin[Cre] transgenic allows targeting the vast majority sensory neurons in
the PNS.[13] To evaluate the representation of neuronal subpopulations within our
dataset, we performed gene set enrichment analysis using a curated list of subtype
markers, described in Barry et al., 2023.[13] As expected, the strongest subtype
enrichment was observed in the DRG, with significantly enriched signatures for non-
peptidergic and peptidergic nociceptors, as well as a mix of low-threshold
mechanoreceptors (LTMRs; include A-field-LTMR, C-LTMR, and AB-LTMR) (Fig 2G).
Across all components a positive trend for each subtype emerged. Significant
enrichment was particularly notable for non-peptidergic nociceptors, A-field-LTMR, and
C-LTMR (Fig 2G, STable 3). These findings indicate that our proximity labelling approach
preserves the molecular diversity of sensory neuron subtypes across their anatomical
domains.

Figure 2. Neuronal proteins are enriched across sensory neuron compartments. A. Protein counts by sample, with "overlap" representing proteins in both Adv^Cre;TurboID^fl/fl and control tissue. B. PCA. C. Heatmap of scaled expression across ion channels in Adv^Cre;TurboID^fl/fl samples. D. Calculated enrichment of neuronal proteins by tissue. E. Comparison of neuronal protein enrichment across tissues. F. Volcano plots highlighted the neuronal enrichment for DEP proteins from D, labelled with neuronal and endogenously biotinylated proteins. G. Gene set enrichment analysis against custom DRG subtype gene lists. Count reflects the number of proteins detected within each gene list, per compartment. GS = gene set. NP = non-peptidergic. PEP = peptidergic. PROP = proprioceptor. LTMR = low threshold mechanoreceptor. LFC = log$_2$ fold change. terminals$_p$ = peripheral terminals (skin); terminals$_c$ = central terminals (SC).

Compartmental sexual dimorphism

To explore sex-specific molecular differences, we analysed differential protein expression between male and female mice across all compartments (SFig 3). With an FDR < 0.05, the sensory neuron proteome appeared largely conserved between sexes, with a few exceptions in abs(LFC) > 1, such as Lgi1 and Sprr1a in the LSC and Brd4 and Palm2 in the peripheral terminals (SFig 3A, STable 4).

To gain broader biological insights, we also conducted gene set enrichment analysis using ranked LFC. This analysis revealed sexually dimorphic regulation of select pathways (SFig 3B). In the peripheral terminals, we see differences in circulatory and vascular pathways, whereas peptide biosynthesis and macromolecule biosynthesis show differences in the somata. These findings suggest localized sex differences in specific biological processes and highlight pathways of potential interest for future mechanistic studies.

Functional signatures show separation by neuronal compartment

To extract the corresponding functional signature, Gene Ontology (GO) pathway analyses were performed against a background of all proteins expressed in the corresponding tissues (Fig 3). While neuronal signatures are detected in each compartment, they also show functional specificity, as expected (Fig 3, STable 5). Primary afferents synapse on secondary order neurons in the dorsal horn of the spinal cord. This corresponds to the enrichment of proteins involved in presynaptic function and synaptic signalling between neurons, as well as vesicle-mediated transport, exocytosis, and other synapse-related terms in the LSC. The DRG contain neuronal somata, and, thus, corresponding functional enrichment matches expectations. Here, we see the enrichment of proteins involved in mRNA processing, post-translational modification, and nuclear structures, in line with the biosynthetic activity expected in neuronal cell bodies.

The sciatic nerve consists of neuronal axons, as well as Schwann cells, fibroblasts, endothelial cells, and immune cells. Its neuronal proteome is enriched for myelination, cytoskeletal organization, and vesicle-mediated transport (STable 5). In the "myelination" terms, we see the presence of nodal and paranodal proteins such as Cntn1 (Contactin-1) and Cntnap1 (Caspr1), as well as more generally expressed neuronal proteins like Ncam1, Nefl, Nefm, and Nefh. We also see proteins more classically associated with myelin itself, such as Plp1, MBP, and MPZ , which also show expression in neuronal subtypes.[18]

Figure 3. Pathway and neuronal subpopulation enrichment across compartments. Schematic highlights the GO over-enrichment analysis per tissue. BP = biological pathway; CC = cellular compartment. (X) term rank by adjusted p-value. % represents the GO term/neuronal enrichment. Please see STable 5 for corresponding data.

To explore functional organization across neuronal compartments, a signed weighted protein co-expression network was constructed on the neuronal proteome (Fig 4). The co-expression network was built using the WGCNA package in R with a soft power threshold of 14 (see, methods). Ten protein co-expression modules were identified (excluding the unassigned "grey" module), each comprising a minimum of 30 proteins (Fig 4A). Of these, 9 modules showed significant correlation with anatomical compartment (FDR < 0.05), while none were associated with sex (Fig 4B).

Within compartments we report significant pathway changes between males and females, as well as a few DEPs (SFig 3). The lack of sex association here suggests this is a relatively weak effect (Fig 4B), consistent with the published findings reporting a mild but present sexual dimorphism in DRG neurons.[13,19]

Specific modules showed clear compartmental specificity. For example, the brown module correlated strongly with the DRG (somata), while the blue module was

associated with central terminals in the spinal cord. Conversely, several modules - including black, yellow, and green - were negatively associated with peripheral terminals, likely indicating a relative depletion of these proteins in distal axonal endings (Fig 4C–D). As expected, most modules exhibited strong alignment with one or more anatomical compartments, reinforcing the functional specialization of the sensory neuron proteome across space (Fig 4D).

Biological pathway analyses of these modules (against a background of all proteins used to build the co-expression network) was then performed (Fig 4E-H, STable 6). The `brown` module, which is positively correlated with the soma, reflects a group of proteins implicated in `translation`, `peptide metabolic process`, and `RNA processing` (Fig 4E). The blue module, which correlated with central terminals in the spinal cord, showed associated with `trans-synaptic signalling`, `regulation of neurotransmitter levels`, and `neurotransmitter transport`, reflecting synaptic functions in the dorsal horn (Fig 4F). The yellow module, which was positively correlated with soma and central terminals, relates to `protein polymerization`, `supramolecular fibre organization` and the `regulation of cellular component size` (Fig 4G).

No modules were positively correlated with the peripheral terminals - possibly due to the lower number of enriched proteins in peripheral terminals - but the most negative correlation is the black module. Although no biological pathways reached statistical significance in this module, cellular components related to `neuronal projection`, `somatodendritic compartment`, `postsynaptic density`, and `neuronal cell body` were among the significantly enriched annotations (Fig 4H).

Figure 4. Weighted co-expression network of the neuronal-enriched proteome. A. Dendrogram
with minimum 30 proteins per module. B. Correlation between modules and traits
(compartment or sex). C. ME (module eigengene) value per sample. D. Module eigengene (ME)
separation by component and module for modules significantly correlated to tissue (FDR < 0.05,
BH correction). E-H. Biological pathway (E-G) and cellular compartment (H) analysis for
modules of interest. E. brown, F. blue, G. yellow, H. black. No biological pathways or cellular

components were significant for the green module. p.adjust = adjusted p-value. qscore = -
$\log_{10}$(p-value). terminals$_p$ = peripheral terminals (skin); terminals$_c$ = central terminals (SC).

TurboID proteomics provides a unique look at terminal endings

Previous -omic analyses of the skin and spinal cord have been unable to resolve the
molecular composition of peripheral and central terminals from primary afferents. We
therefore leveraged the cell type and spatial precision of the TurboID approach to
functionally characterize these neuronal compartments (Fig 5, SFig 4). To explore
protein interaction and functional organization, large-scale protein–protein interaction
(PPI) networks were constructed using STRING.[20] A Girvan-Newman fast greedy
algorithm was then used to generate community clusters[21,22], and the functional
enrichment of each cluster was calculated.

In the peripheral terminals (skin), the proteome revealed a diverse and functionally rich
landscape, spanning protein transport, metabolic pathways, and signal transduction
(Fig 5). A wide array of protein classes was detected, including SNARE complex
components such as Syntaxin-7 and Snap25, as well as chaperones and ubiquitin-
related proteins, such as Usp7 and heat shock proteins. Notably, even pain-associated
ion channels (Trpv1, Scn10a, and Scn11a) were detected in clusters annotated for
membrane trafficking and exocytosis. A full breakdown of each cluster enrichment is
available in STable 7.

A similar pattern was observed in the central terminals (spinal cord; LSC), where
clustering revealed extensive functional heterogeneity (SFig 4, STable 8). Enriched
clusters included pathways involved in signal transduction, endosomal processing, and
membrane trafficking—hallmark processes of synaptic signalling and
neurotransmission (see Stable 8 for a functional breakdown). Together, these analyses
demonstrate that proximity labelling via TurboID enables unprecedented access to the
proteomic architecture of otherwise inaccessible neuronal compartments. This
approach holds considerable promise for extracting functional, spatially resolved
proteomes in defined cell types.

TurboID proteomics distinguishes neuronal proteomes from bulk tissue backgrounds

To contextualize the advantages of TurboID-based proximity labelling, we compared our
compartment-resolved neuronal proteomes to conventional bulk tissue proteomics.
We harvested the skin, SCN, DRG and SC and determined the proteome using the same
mass spectrometry-based workflow as applied to the TurboID samples above (SFig 5,
STable 9). As expected[23,24], we detect several thousands (7000-9000) protein groups per
tissue (SFig 5A), with samples clustering by tissue (SFig 5B). Consistent with earlier
studies[23,24], we identified many neuronal proteins in bulk datasets. Overall, these bulk
tissue proteomes exhibited a large overlap with neuronal proteomes defined by our

TurboID approach (Fig. 1-4), while dozens of proteins were identified uniquely in each
neuronal proteomes (SFig 5D, STable 10).

TurboID proteomics depletes non-targetted cell type proteomes

The TurboID workflow enables us to selectively define the neuronal proteome within
heterogeneous tissue (Fig 2-5, STable 10). To confirm the enrichment of sensory
neurons under the Advillin driver, we also investigated the depletion of immune
subtypes across compartments (SFig 6). ImmuCellAI[25] was used to calculate an
"immune infiltration score", which reflects a composite GSEA enrichment across
immune subtypes. In every compartment, we see a reduced immune signature
(compared to bulk tissue samples), further validating our approach (SFig 6).

This highlights two key advantages of TurboID-based neuronal enrichment: First, it
facilitates the discrimination of proteins expressed in neurons versus non-neuronal
cells within the same tissue. Second, TurboID enhanced detection sensitivity in tissues
with sparse neuronal content, such as in the skin. In these regions, the detection of
sensory neuron-specific proteins — particularly ion channels — has been challenging
and largely unsuccessful using bulk proteomics.[23,26] In contrast, TurboID-based profiling
recovered functionally important signatures in peripheral terminals, including
prominent drug targets such as Scn10a, Scn11a, and Trpv1, providing molecular access
to previously inaccessible compartments (Fig 5, SFig 5D).

Together, these data underscore the power of TurboID-based proteomics to resolve cell
type-specific proteomes within complex tissues, particularly for rare or spatially
restricted compartments. Importantly, this strategy opens a path to study the proteome
dynamics of afferent terminals in disease models and in response to pharmacological
interventions. Beyond sensory neurons, our TurboID[fl/fl] workflow can be readily adapted
with other driver lines to define functionally relevant, compartment-specific proteomes
across diverse tissues and cell types.

Figure 5. Community clustering of the peripheral terminal proteome, with edges denoting high confidence protein-protein interactions. Protein nodes are coloured by cluster, and clusters are described by functional summaries with proteins of interest highlighted. Full functional enrichments are available in STable 7.

TurboID proteomics reveals neuronal-specific changes in response to oxaliplatin treatment

Next, we aimed at showcasing the utility of our TurboID workflow to assess neuronal proteome changes in a pharmacological context. For this, we turned to oxaliplatin, a chemotherapeutic agent. Chemotherapy-induced peripheral neuropathy (CIPN) is a common and often painful condition affecting the peripheral nervous system, known to impact neuronal and non-neuronal cells in the DRG.[12] In adherence to the 3Rs of animal research, we employed a DRG explant model to explore oxaliplatin-induced changes in the neuronal somata (Fig 6, SFig 7-8, STables 11-16).

Mirroring the compartmental proteomes above, we determined the bulk proteome of DRG explants from wildtype mice and compared these data with neuronal proteomes obtained from Adv^Cre;TurboID^fl/fl mice. In the bulk proteome of wildtype mice we identified ~7200 protein groups (SFig 7A, STable 11), with clustering driven by oxaliplatin concentration (SFig 7B-C). Consistent with earlier studies, oxaliplatin exposure led to

widespread proteome changes, including alterations in known neuronal proteins (SFig 7D-E, STable 12). When comparing differentially expressed proteins (DEPs) from our DRG explant data to previous work *in vivo*[27], we see a strong correlation between the general expression of these proteins (SFig 7F). We next employed a rank-rank hypergeometric overlap test to compare the differential protein expression from their *in vivo* data to ours (SFig 7G). This shows good concordance for protein changes across both datasets, suggesting similar regulation by oxaliplatin and validating our *in vitro* approach.

In DRG explants from Adv^Cre^;TurboID^fl/fl^ mice, we observed specific alterations of the neuronal explant proteome in response to oxaliplatin (Fig 6, SFig 8, STable 13). As expected, Adv^Cre^;TurboID^fl/fl^ explants exhibited prominent protein enrichment compared to TurboID^fl/fl^ controls (SFig 8A). Samples clustered by genetics in the first principal component (PC), with slight separation by drug treatment in the second PC (SFig 8B). In both oxaliplatin-treated (SFig 8C, left) and vehicle-treated (SFig 8C, right) explants, Adv^Cre^;TurboID^fl/fl^ samples (vs TurboID^fl/fl^) showed robust enrichment of neuronal proteins with ~ 2500 protein groups per condition (SFig 8D, STable 14).

Within the neuronal explant proteome, we detected a substantial number of DEPs after oxaliplatin treatment (Fig 6A, STable 15). Notably, these differences were only partially reflected in the bulk explant proteome (Fig 6B). Even so, we saw a high rank-rank hypergeometric overlap between the bulk explant and neuronal proteomes after oxaliplatin treatment (Fig 6C), and a high semantic similarity across DEPs (Fig 6D).

Next, we assessed DEP-associated functions. After oxaliplatin treatment, pathways related to amino acid transport and cell adhesion were downregulated in the neuronal explant proteome (Fig 6E, STable 16). To further dissect these pathways, we constructed a corresponding PPI network based on pathway-associated DEPs (Fig 6F). This revealed prominent downregulation across multiple functional modules (Fig 6Fi). While some of these DEPs were detected in both bulk and neuronal explant proteomes, others were uniquely regulated in the neuronal proteome (Fig 6Fii). The latter includes Trpv1, Scn11a, and Calca, as well as the central module with Itgb1, Tin2, and Rock2. These findings illustrate the enhanced resolution of cell type-specific proteomics using TurboID, uncovering molecular changes that may be masked in bulk tissue analyses.

In summary, targeted cellular proteomics can uncover relevant proteomic differences across compartments and pharmacological treatments, offering a powerful tool to dissect disease mechanisms and treatment effects in complex tissues.

Figure 6. Oxaliplatin-treated Adv[Cre];TurboID[fl/fl] explants highlight neuronal dysregulation. A.
Volcano plot for oxaliplatin vs vehicle, neuronal proteome (Adv[Cre];TurboID[fl/fl]). B. DEP overlap
(oxaliplatin vs vehicle) between bulk (WT, grey) and neuronal (Adv[Cre];TurboID[fl/fl], orange) explant
proteomes. C. Correlation between LFC for overlapping DEPs in B. D. Semantic similarity
between neuronal and bulk DEPs. E. Neuronal biological pathway downregulation after
oxaliplatin treatment, highlighting protein counts which overlap across pathways. F. STRING
protein-protein interaction network for DEPs in regulated pathways from E. i, LFC from neuronal

proteome. ii, Specificity to neuronal analysis. DEPs detected only in Adv[Cre];TurboID[fl/fl] are in

orange, those overlapping with the bulk analysis are grey.

Discussion

As a generalized proof-of-concept, we demonstrate how a TurboID transgenic mouse

enables cell-type-specific proteomic labelling across anatomically distinct

compartments. When paired with Advillin[Cre], TurboID expression is targeted to primary

sensory neurons of DRG, resulting in robust, neuron-enriched biotinylation. Using DIA-

PASEF mass spectrometry, we profiled compartment-specific DRG proteomes and

uncovered enrichment of neuron-specific proteins and pathways. These included

canonical markers and functionally relevant processes, such as translation in the

somata and neuron-to-neuron synapse functions in the central terminals of the spinal

cord.

This spatial resolution is particularly significant in the peripheral terminals of DRG

neurons in the skin. These nerve endings are typically characterized by

immunohistochemistry and have remained largely inaccessible to unbiased omics

approaches. Neither our bulk proteomic data nor previous high-coverage DIA-MS

datasets from mouse skin[23] or human skin[26] detected key sensory neuron markers. In

bulk analyses, low-abundance neuronal signatures are often masked by dominant cell

types. Using TurboID-based proteomics, we were able to detect key markers and drug

targets such as Trpv1, Scn10a, and Scn11a in the peripheral terminals.

While single-cell and single-nucleus RNA sequencing are valuable for cell-type-resolved

transcriptomes[28,29], they are limited to somata within the DRG, lack proteomic

resolution, and do not capture distal compartments. [28,29] The expansion of spatial

sequencing helps address many of these issues, although recent spatial

transcriptomics of the sciatic nerve have not resolved neuron-specific clusters[30], and to

our knowledge, a high resolution spatial-seq dataset including peripheral terminals in

the skin is still missing. As these methods mature, they will complement our data well.

At present, however, our TurboID-based proteomic atlas offers a unique spatial profile

spanning the full extent of mouse sensory neurons. While not a main focus of this study,

we were also able to probe neuronal sex differences across compartments, expanding

on previously reported sexual dimorphism in mouse and humans.[13,19,23,31] Mirroring

recent transcriptomic data across sensory neuron subtypes, we see some, but not

extensive differences between male and female neurons.[13] Few proteins were sexually

dimorphic within compartments, including Kct2 in the soma and Brd4 in the peripheral

terminals (both enriched in females), as well as Serpina3k in the axon (enriched in

males). Previous bulk proteomic analyses of mouse sciatic nerve and paw have also

identified Serpina3k enrichment in males[23], although no DEPs from the paw were shared

with their bulk data, suggesting a unique signature from the peripheral terminals versus

whole skin preparations.[23] In the central terminals of the spinal cord, we see an

enrichment of Lgi1 and Sprr1a in males. Sprr1a is an injury marker with reported male-
biased mRNA expression in DRG[13], though the functional consequences remain to be
defined. Lgi1 has recently been implicated in pain sensitivity[25], and future studies are
needed to investigate whether the proteomic differences observed here translate to
phenotypic effects.

To highlight how this resource could be exploited to study functional changes in a
clinically-relevant disease context, we next studied neuronal changes in a DRG explant
model of CIPN. Exposure to oxaliplatin resulted in widespread proteomic alterations in
both bulk and neuron-enriched samples. While the sets of differentially expressed
proteins (DEPs) varied between bulk and TurboID-labelled samples, semantic similarity
of enriched pathways was high, indicating shared biological responses. A major benefit
of the Adv^{Cre};TurboID$^{fl/fl}$ approach in explants is the cell type specificity, highlighting
disruption to organic anion and amino acid transport pathways in neurons. This
contrasts the more generalized effect of oxaliplatin seen in the whole DRG proteome,
where we lose the cell type resolution.

Transporter disruptions have widely been reported in the context of CIPN.[12,32–36] Organic
cation transporters, such as Oct2, have been implicated in oxaliplatin uptake by
sensory neurons, and their activity contributes to neurotoxicity.[12,36] Similarly, organic
anion-transporting polypeptides (OATPs) affect CIPN pathophysiology.[32,37,38] Impaired
transporter function can lead to drug accumulation in DRG, contributing to
mitochondrial dysfunction, oxidative stress, and neuroinflammation, all of which
underlie CIPN pathogenesis.[12] In our study, several transporters were downregulated in
response to oxaliplatin treatment. Some changes are evident across neuronal and bulk
proteomes (e.g. Slc7a14, Slc2a3), while the dysregulation of others (e.g. Slc1a4, Slc3a2,
Slc33a1) could only be detected when studying the neuronal proteome. Pursuing these
transport pathways may provide novel strategies to mitigate CIPN while preserving the
anticancer efficacy of chemotherapy, with promising attempts to target transporters
already in development.[37]

Finally, our findings highlight the practicality of the DRG explant model for rapid *in vitro*
screening of neurotoxic effects. It also reinforces the added value of compartment-
specific proteomics in detecting functionally relevant molecular changes. When paired
with conventional bulk analyses, our approach offers a multi-layered view of the
proteomic landscape in response to drug treatment and in disease models,
complementing and extending previous *in vivo* work.[27]

Limitations

This study provides a quantitative, compartmental proteomic atlas of DRG sensory
neurons and serves as a valuable reference for the field. However, it is likely not
exhaustive. Detection of low-abundance proteins may be further improved through

advances in mass spectrometry sensitivity, biotinylation efficiency, and sample
preparation protocols. As expected, fewer protein identifications were obtained in
peripheral terminals and axons - a likely reflection of tissue composition and neuronal
content - resulting in a relative bias toward somata and central terminals, where
neuronal protein density is higher. While these results represent a robust starting point,
future studies with deeper coverage and within-compartment comparisons under
various biological conditions will be critical for advancing our knowledge about the
molecular features of terminal regions.

Future directions

A key advantage of using TurboID transgenic mice is the opportunity to interrogate
otherwise inaccessible cellular compartments. Prior work in both mice and humans has
shown alterations in intraepidermal nerve fibre density across painful conditions.[1,39]
Pairing our compartmental proteome with *in vivo* functional testing (e.g. oxaliplatin
administration) will provide a rich understanding of the underlying molecular changes
and their correlation with behavioural outcomes in future studies.

Another major benefit lies in the fact that Cre-based recombination allows flexible
customisation. Pseudobulk analyses are frequently used in scRNA-seq (and spatial)
transcriptomics to compare within cell-type changes across conditions. Paired to
adequate transgenic drivers and downstream DIA-MS, TurboID[fl/fl] can be used to
generate a corresponding proteome for multi-omic analyses within cell types. In the
context of sensory neurons, this can be subtype targeting (e.g. Mrgprd[Cre] or Th[Cre]) or
injury-specific targeting (e.g. Atf3[Cre]).[14,40] More broadly, our approach could be extended
to diverse cell types, such as fibroblasts in the skin, satellite glia in the DRG, or
interneurons in the dorsal horn (to name a few).

Conclusions

In this study, we generated a TurboID[fl/fl] transgenic mouse, using it to systematically
profile the proteomes of DRG sensory neuron compartments. Spanning from peripheral
to central terminals, this study presents a proteomic atlas across DRG sensory neuron
compartments. Our findings reveal compartment-specific protein distributions that
reflect specialized neuronal functions. Furthermore, we report previously unknown
neuron-specific changes in an explant model of chemotherapy-induced peripheral
neuropathy. Taken together, our methodological workflow and obtained data serve as a
highly valuable resource for studying neuronal specialization and proteome dynamics in
the peripheral nervous system.

Code Availability

All code is available at https://github.com/aliibarry/turboID.

Data Availability

The mass spectrometry proteomics data, including metadata, fasta, and gene group
matrices for each dataset have been deposited to the ProteomeXchange Consortium
via the PRIDE[41] partner repository with the dataset identifier PXD062198. Processed
data are also available on github (https://github.com/aliibarry/turboID/data/).

For queries regarding the TurboID mice please contact Nils Brose
(brose@MPINAT.mpg.de) or Noa Lipstein (lipstein@FMP-berlin.de).

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Author Contributions

Method optimization, sample preparation and data collection were performed by JRS
with input from FX, DVG, and MS. Analyses were performed by AMB, with input from JRS,
FX, and MS. Mass spectrometry was performed by FX with input from JRS and DGV.
Conceptualization and funding acquisition by MS with input from DGV. Oxaliplatin
experiments were performed by JRS and TH. TurboID$^{fl/fl}$ mice were conceived, designed,
generated and validated by NB, FB, DS, and NL. Manuscript was written by AMB, JRS,
and MS. All authors approved the submitted version.

Declaration of interests

556 The authors declare no conflicts of interest related to this work. NL is a scientific
557 advisory board member of TRACE Neuroscience inc. MS and DGV maintain an ongoing
558 scientific collaboration with Bruker, but this collaboration had no influence on the
559 content of the manuscript.

Methodology

Generation of TurboID mice (ROSA26$^{em1(TurboID)Bros}$) using CRISPR/Cas9

The conditional TurboID knock-in (cKI) mouse line was generated by site-directed CRISPR/Cas9 mutagenesis in the ROSA26 locus. Superovulated FVB/J females were mated with FVB/J males and fertilized eggs collected. In-house produced CRISPR reagents (hCas9_mRNA, sgRNAs, preformed Cas9_sgRNA-RNP complexes and the homology-directed repair fragment (HDR fragment)) were microinjected into the pronucleus and the cytoplasm of pronuclear-stage zygotes using an Eppendorf Femtojet. Importantly, all nucleotide-based CRISPR/Cas9 reagents (sgRNAs and hCAS9_mRNA) were used as RNA molecules and were not plasmid-encoded, reducing the probability of off-target effects due to the short lifetime of RNA-based reagents.[42,43] The TurboID-containing HDR fragment was generated by subcloning a V5-tag-TurboID sequence into the Ai14-ROSA26-based plasmid backbone[44] and isolated as a linear DNA fragment by SacI/AgeI restriction enzyme double digestion. Two sgRNAs targeting the ROSA26 Locus were selected using the guide RNA selection tool CRISPOR.[45,46] The correct locus-specific insertion of the HDR fragment was confirmed by localization PCRs with primers upstream and downstream of the HDR sequence. Extended details, including full sequences and genotyping strategy, are available in the Supplemental Methods.

Animal husbandry

Adult male and female TurboID$^{fl/fl}$ control/ Adv^{Cre};TurboID$^{fl/fl}$ mice (12.5-26 weeks old for compartmental proteome, 57-71 weeks old for oxaliplatin part) and in-house bred wild type C57BL/6J male mice (60-66 weeks old) were kept in a 12/12h day/night cycle with *ad libitum* access to food and water in an open-cage environment. TurboID$^{fl/fl}$ control and Adv^{Cre};TurboID$^{fl/fl}$ mice received biotin-enriched drinking water (0.22 mg/mL, prepared freshly every day) for 7 consecutive days and on day 8 the mice were sacrificed.

The animal experiments were conducted in accordance with ethical guidelines and were approved by the Institutional Animal Care and Use Committee (IACUC) at the University of Vienna and the Austrian Ministry for Education, Science, and Research (BMBWF; license number 2021-0.138.925 amended by 2024-0.896.105). All procedures complied with the international ARRIVE guidelines and adhered to the European Union Directive 2010/63/EU on the protection of animals used for scientific purposes. No mice were removed from the study before their expected endpoint.

The compartmental proteome was generated from pooled samples (4-6 per samples per group per tissue) from the following mice: TurboID$^{fl/fl}$ control = 29M + 29F and Adv^{Cre};TurboID$^{fl/fl}$ = 19M + 27F.

Unilateral mouse DRG were used for one explant culture stimulated with Oxaliplatin,
the DRG of the other side for the Vehicle condition in parallel. For the enrichment of
biotinylated proteins in Adv^{Cre};TurboID^{fl/fl}, 4 samples (3M, 1F) were prepared for each
condition, with multiple mice pooled for each samples. In total, 10 males and 4 females
were used per condition (3-4 mice per sample).

Tissue dissection

Mice were sacrificed via CO_2-inhalation and rapidly decapitated. The vertebral spine
was removed for subsequent isolation of the dorsal part of the lumbar 3-5 spinal cord
(SC) and lumbar 3-5 dorsal root ganglia (DRG). In parallel the sciatic nerves (SCN) of
both sides were dissected. The plantar paw skin (dermis and epidermis, without muscle
tissue) was removed by a 4 mm skin punch. Skin samples were immediately snap-
frozen in liquid N_2, all other tissues were washed in ice-cold 1× PBS to remove blood
residuals before snap-freezing in liquid N_2. All tissues remained at -80°C until further
use.

Protein extraction

All steps are done at room temperature (RT) except otherwise mentioned.

DRG: Tissue homogenisation was done in lysis buffer (LB; 2% SDS, 0.1 M tris, 5%
glycerol, 0.01M DTT, 1× complete protease inhibitor cocktail) with 3x 10 strokes of a
pestle/potter. To facilitate lysis the homogenate was sonicated using a Bioruptor Pico
(10x 30 s ON/30 s OFF, 4°C, low frequency, all sonification steps mentioned below are
done with the same settings, only cycle number varies) and further incubated at 70°C,
for 15' min with 1500 rpm (this and all other incubation with agitation were done in a
Thermomixer C, Eppendorf). The remaining cell debris was pelleted by centrifugation
(10', 10000 ×g) and the supernatant was saved. All replicates of one condition were
pooled and the concentration determined at 280 nm with a NanoPhotometer N60
(Implen, München, Deutschland).

SCN: Protocol was performed as above, except the centrifugation was followed by
another 5moin-long centrifugation (RT, 10000×g) to further clear the lysate of lipid
clouds. Before pooling and concentration measurement an acetone precipitation was
done (see below).

SC: The DRG protocol was modified in that the pestle/potter step was omitted, instead
the SC was cut into 3 smaller pieces and transferred to 2 mL LoBind tubes (prefilled with
500 µL of LB) for sonication Pico for 15 cycles. Other steps as for DRG, but before
pooling and concentration measurement an acetone precipitation was done (see
below).

Paw skin: the skin punches were cut into 4 smaller pieces, transferred into a potter
prefilled with LB and homogenized with help of 3× 10 strokes of a pestle/potter. Lysis of
tissue was facilitated by incubation for 20 min, at 70°C and 1500 rpm, followed by

sonification for 15 cycles. The lysate was cleared off cell debris and excessive lipids by
2× centrifugation 10 min, 10000×g) and the supernatant was saved. Before pooling and
concentration measurement an acetone precipitation was done (see below).

For the SCN, SC and paw skin an acetone precipitation was done to reduce the amount
of lipids in the samples: 5× sample volume of pre-cooled (-20°C) 100% acetone was
added to each sample and the tube was shortly inverted to ensure thorough mixing. If
needed, samples were split into 2× 2 mL tubes to allow for 5× addition of sample
volume. Precipitation of proteins was allowed for 3-3.5 h at -20°C. Next the precipitates
were pelleted by centrifugation for 30 min and 14000×g. Pellets were washed with 1.5
mL pre-chilled (4°C) 80% ethanol and the centrifugation repeated as above. After
removal of the ethanol, pellets were air-dried for 20-30 min. Proteins were reconstituted
in LB by incubation for 15 min, 70°C, 1500 rpm. If were not fully solubilized, the
incubation was repeated and LB volume increased until fully solubilized. Total protein
concentration of pooled samples was determined with a NanoPhotometer at 280 nm.

Streptavidin-mediated enrichment of biotinylated proteins

Streptavidin-affinity purification protocol is based on the high stringency protocol
described in Go et al., 2021 with modifications. The required amount of streptavidin
beads (Sera-Mag SpeedBead Blocked Streptavidin Particles, Cytiva, 83 µL bead slurry/
1000 µg input) was transferred into LoBind protein tubes, washed twice with 500 µL LB
without SDS, and resuspended in the same buffer. Protein lysates (input amount: DRG
1000 µg, SCN 600 µg, SC 800 µg, skin 2000 µg) were prepared in a way that the final SDS
concentration was 0.5% in a final volume of 1500 µL and incubated with the beads at
4°C for 3.5 h on an overhead shaker. Unless otherwise stated, all subsequent steps
were performed at RT.

Following incubation, samples were placed on a magnetic rack for 2 min. Beads were
then subjected to a series of washes, each performed with 1 mL of buffer, followed by
overhead shaking for 1 min each (Grant-Bio PTR-35 Multi-Rotator)and incubation on the
magnetic rack for 2 min until the solution cleared. The washes included one step with
2% SDS Wash Buffer (2% SDS, 25 mM Tris), two washes (three for skin samples) with
modified RIPA buffer (50 mM Tris, 150 mM NaCl, 1.5 mM $MgCl_2$, 1% Triton X-100, 1 mM
EDTA), and three washes with 50 mM ammonium bicarbonate (ABC, pH 8). After the first
ABC wash, the beads were transferred to a new 1.5 mL LoBind tube in 500 µL ABC
buffer, followed by the addition of another 500 µL ABC before finalizing the washes.

For protein reduction, 135 µL of 5 mM dithiothreitol (DTT) in 50 mM ABC containing 0.8
mM biotin was added to each sample and incubated at 60°C for 30 min at 1000 rpm.
Samples were then cooled to RT by briefly placing them on ice. Alkylation was
performed by adding iodoacetamide (IAA) to a final concentration of 20 mM (15 µL of
200 mM IAA), followed by incubation for 30 min in the dark under constant agitation. The
reaction was quenched by adding 1M DTT to a final concentration of 5 mM and

incubating for 15 min with 3× 5 s vortexing steps in between. Proteins were digested for

17 h at 37°C and 950 rpm using 1.5 µg Trypsin/LysC per 1000 µg input. After the digest,

peptides were eluted by placing the tubes on a magnetic rack for 2 min and transferring

the supernatant to fresh 2 mL LoBind tubes. The beads were further washed with 20 µL

MS-grade water (5 min, 500 rpm), and the wash was pooled with the initial eluate.

Peptide concentrations were measured at 205 nm to calculate the required amount of

SP3-beads (1:1 mix of Sera-Mag SpeedBead beads, 10-15 µL bead slurry to 40-50 µg

peptides). To clean-up the peptides, peptide solution was split into two 2 mL tubes, the

required bead amount and 100% ACN was added to reach 95% ACN. Samples were

incubated for 8 min and 500 rpm, followed by a 2-min incubation on a magnetic rack.

The ACN was discarded, and beads were rinsed with 400 µL 100% ACN, incubated for 2

min on a magnetic rack, and the ACN was discarded again. Peptides were reconstituted

in 20-25 µL MS-grade water, incubated for 5 min with shaking at 500 rpm, followed by

another 2-min incubation on a magnetic rack. The supernatants of previously split

samples were pooled, centrifuged at 17000×g for 2 min, placed on a magnetic rack, and

the final supernatant was transferred to a glass vial. Peptide concentrations were

measured at 205 nm with a NanoPhotometer, and samples were acidified to a final

concentration of 0.1% formic acid (FA). The samples were stored at -20°C until LC-MS

measurement.

DRG explants preparation

From both Adv[Cre];TurboID[fl/fl] and TurboID[fl/fl] mice (fed with biotin-enriched water for 7

days, see above) 20 DRG per side were isolated and collected separately by side in two

tubes containing DMEM/F-12 at RT. The DRG of one side were plated in 50 µL volume in

to a well of a 24-well plate and incubated for 10 min at 37°C, 5% CO_2. In the meantime,

the oxaliplatin (100 µM) and/or vehicle (H_2O, Veh) was mixed with DMEM +10% horse

serum (HS) +100 ng/mL nerve growth factor (NGF) + 50 µM biotin and 500 µL/well were

carefully added. The drug/Veh was incubated for 48 h at 37°C, 5% CO_2.

To find an effective oxaliplatin concentration in line with previous reports[47], DRG explant

cultures were prepared in a similar way from wild type mice. DRG from one mouse were

separated into 4 wells for 4 conditions Veh, 25 µM, 50 µM and 100 µM oxaliplatin in

DMEM+HS+NGF. The drug/Veh was incubated for 48 h at 37°C, 5% CO_2.

Sample preparation of DRG explants

The protein extraction was done similarly for both WT and TurboID DRG explants. All

DRG of one well were transferred to an Eppendorf tube and washed twice with ice-cold

1× PBS by inverting the tube 10×. After discarding all PBS, 300 µL LB was added and

DRG explants were sonicated for 10 cycles and further incubated at 70°C, for 15 min

and 1500 rpm. Remaining cell debris was pelleted by centrifugation (5', 10000×g) and

the supernatant was saved. The total protein concentration was determined at 280nm

with a NanoPhotometer N60. The TurboID samples underwent the above described

streptavidin-affinity purification to enrich for biotinylated proteins and preparation of
peptides (input 725 µg).

For the WT samples a modified version of the single-pot, solid-phase-enhanced sample
preparation (SP3) method[48] was employed to digest proteins and clean the peptides. In
brief, 10 µL of pre-mixed Sera-Mag SpeedBead beads were added to 50 µg of protein. To
facilitate protein binding to the beads, an equal volume of absolute ethanol was added,
followed by incubation for 5 min at 24°C and 1000 rpm. Beads were collected on a
magnetic rack for 2 min, the supernatant was discarded, and the beads were washed
thrice with 500 µL of 80% ethanol. Subsequently, the beads were resuspended in 50 µL
of 50 mM ABC, pH 8 containing 2 µg of Trypsin/Lys-C and incubated overnight at 37°C
and at 950 rpm. Following protein digestion, ACN was added to each sample to reach a
final concentration of 95%. After an 8-min incubation, beads were collected on a
magnetic rack for 2 min. The supernatant was discarded, and the beads were washed
with 900 µL of ACN. For peptide elution, the remaining ACN was allowed to evaporate
for 2-3 min and the beads were resuspended in 40 µL of LC-MS grade water. Peptide
concentrations were measured at 205nm using a NanoPhotometer N60. Finally, the
peptide samples were acidified with FA to a concentration of 0.1% and stored at -20°C
until LC-MS/MS analysis.

Gel-electrophoresis & Western blot

All lysates from TurboID tissue that were going to be used as input samples for the
streptavidin-affinity enrichment were checked by WB for the biotinylation success. The
samples were prepared by dissolving them in 4× LDS Sample Buffer supplemented with
4× Reducing Agent to reach final 1×concentrations followed by heating to 70°C for 10
min. For protein separation, samples were loaded onto pre-cast NuPAGE® Bis-Tris 4–
12% gradient gels, the NuPAGE® (Life Technologies) electrophoresis chambers filled
with 1× NuPAGE® MOPS buffer and the gels run for 40-50 min at 200 V.

Following electrophoresis, proteins were transferred onto PVDF membranes using the
iBlot2® Gel Transfer Device (Life Technologies, program #0, 7 min). To block non-
specific binding sites, membranes were incubated with 5% milk in PBS for 45 min at RT.
Subsequently, membranes were incubated with primary antibodies (diluted in 1%
milk/PBS; Stable 1) for 1.5 h, followed by an overnight incubation at 4°C with an
additional 1:4 dilution. Membranes were washed with PBS (three short washes followed
by three 5-min washes) before incubation with the respective secondary antibodies
(diluted in 1% milk/PBS; see Stable 1) for 1 h. After washing with PBS as described
before, immunolabeled proteins were detected using near-infrared fluorescence
imaging (BioRAD ChemiDoc MP).

LC MS acquisition

Nanoflow reversed-phase liquid chromatography (Nano-RPLC) was performed using
NanoElute2 systems (Bruker Daltonik, Bremen, Germany) and coupled to a timsTOF HT

mass spectrometer (Bruker Daltonik, Bremen, Germany) via a CaptiveSpray ion source.
The mobile phase consisted of solvent A (100% water, 0.1% formic acid) and solvent B
(100% acetonitrile, 0.1% formic acid).

Peptides were first loaded onto a C18 trap column (1 mm × 5 mm, ThermoFisher) before
being separated on an Aurora™ ULTIMATE analytical column (25 cm × 75 μm) packed
with 1.6 μm C18 particles (IonOpticks, Fitzroy, Australia) over a 60-minute gradient. The
flow rate was maintained at 250 nL/min, except for the final 7 minutes, where it was
increased to 400 nL/min. The gradient of mobile phase B was adjusted as follows: an
initial linear increase from 2% to 15% over 30 minutes, followed by a rise to 35% over 22
minutes, and a sharp increase to 85% within 0.5 minutes. The flow rate was then
adjusted to 400 nL/min in 0.5 minutes and held for 7 minutes to ensure complete
elution of hydrophobic peptides.

Data acquisition was performed in data-independent acquisition (DIA) mode using
parallel accumulation serial fragmentation (PASEF). Precursors with an m/z range of
350–1100 were analyzed across 12 MS/MS ramps in a cycle, with either two or three
quadrupole switches per ramp. Each cycle contained 30 MS/MS windows within a 0.65–
1.30 (1/k0) ion mobility range, using a fixed isolation window of 25 Da per step. The
acquisition time per DIA-PASEF ramp was set to 100 ms, resulting in a total cycle time of
approximately 1.38 seconds. Collision energy was ramped linearly from 65 eV at 1/k0 =
1.6 to 20 eV at 1/k0 = 0.6.

Spectral deconvolution with DIA-NN

DIA-NN 1.8.2[49] was used to process DIA-PASEF data in library-free mode. For the
TurboID and WT samples, each tissue and, in case of Adv[Cre];TurboID[fl/fl] and TurboID[fl/fl]
genotypes were processed in a separate analysis step using the same spectral library.
The spectral library was predicted from a Mus musculus proteome database from
UniProt containing 17,070 protein entries (downloaded on 2021-07-08), in case of the
TurboID samples it included in addition streptavidin (P22629|SAV_STRAV) and TurboID
sequence.[4] *In silico* digestion was performed using Trypsin/P, allowing up to two missed
cleavages. A maximum of three variable modifications was allowed per peptide
including methionine oxidation, and N-terminal acetylation, while Cysteine
carbamidomethylation was set as a fixed modification. Peptide length for the search
ranged from 5 to 52 amino acids. Precursor charge was set to 2-4. In accordance with
the DIA-PASEF acquisition method, the m/z ranges were set to 350–1100 for precursor
ions and 100–1700 for fragment ions. Mass accuracy was chosen automatically using
the first run in the experiment. Protein inference was performed using unique genes. For
neural network classifier the single-pass mode was chosen. Quantification was carried
out using RT-dependent cross-run normalization and the Quant UMS (high precision)
option. The main DIA-NN output was processed using the R package DiaNN to extract
MaxLFQ intensities[49,50] for all identified protein groups, applying a q-value threshold of

<0.01 at both the precursor and gene group levels. Matrices were merged by protein

group were applicable for downstream analyses.

Neuronal protein enrichments

Protein enrichments were calculated within each tissue/neuronal component prior to

merging. First, proteins were filtered for distinct gene symbols, keeping the first instance

where matches exist. Any protein detected in ≥75% of Adv[Cre];TurboID[fl/fl] samples for a

given tissue, and no control sample was considered enriched (type "A"). This comprised

the majority of enrichments. For proteins in ≥75% of Adv[Cre];TurboID[fl/fl] samples from any

tissue, and any amount of control underwent differential expression testing using a

standard limma framework (`lmFit` + `eBayes`).[51] Proteins were modelling by a

grouped factor for Tissue and TurboID Status, and contrasted within tissue between

TurboID and Control samples (ie. Paw.Turbo – Paw.Control). Proteins were considered

enriched when FDR < 0.05, and LFC > 1 ("DEP").

For explant analyses, we used the cross-tissue enrichments described above as a

reference for neuronal enrichment, instead of setting a 75% threshold (ie. explant data

were first filtered for neuronal enrichments using this reference list). Any proteins

detected in the turbo samples and not control (type "Specific"), as well as any proteins

that then hit a threshold of FDR < 0.05, and LFC > 1 (type "DEP") were considered

enriched in explant neurons.

Pathway analyses

GO enrichments were performed on enriched proteins and differentially expressed

proteins using the `enrichGO` function from the ClusterProfiler and org.Mm.eg.db

libraries in R[52]. Both cellular component (CC) and biological pathways (BP) were

extracted, with an adjusted p-value cutoff of 0.01 (Benjamini-Hochberg corrected). To

account for redundancy in the GO terms, outputs were slimmed using default settings:

`clusterProfiler::simplify(ego, cutoff=0.7, by="p.adjust", select_fun=min, measure =

'Wang')`.

For pathway analyses of TurboID-enriched proteins, a background of all proteins

detected across the four tissue wildtype whole cell lysates + proteins detected with

TurboID samples – this represents the entire universe of proteins which could have been

enriched in neurons using the TurboID approach. For pathway analyses of differentially

expressed proteins, a background of all proteins tested within the comparison was

used. For pathway analyses of co-expression modules (below), module genes were

compared to the universe of neuronal-enriched proteins.

Subtype enrichment

Neuronal subtype enrichment was calculated using an Overrepresentation Analysis

(ORA) on previously published custom-made gene lists from the mouse DRG

transcriptomics[13,14]. Briefly, custom gene enrichment lists for DRG neuronal subtypes

were calculated from Zheng et al. (2019) using DESeq2[14,53]. ORA was then performed on
the neuronal enriched proteins from each tissue using `Clusterprofiler::enricher`[52].
For each tissue, the background was defined as any protein detected in whole cell
lysate or TurboID samples, and the custom enrichment lists were filtered for proteins in
the pre-defined background prior to ORA. `Clusterprofiler::enricher` does not allow
the definition of both a custom background universe and custom gene set list (instead
using only proteins defined in the custom gene sets as a background). To overcome this,
the background proteins were defined as an additional gene set, and the gene set size
was capped at 2000 to exclude this set from a direct enrichment analyses while
maintaining them in the background. An FDR < 0.05 was considered significantly
enriched, with an FDR =1 set for plotting.

Immune cell enrichment

The depletion of immune cells was determined by the "infiltration score" from the
`ImmuCellAI` package (0.1.0), a GSEA-based approach to query immune signatures.[25]
Protein names were converted to human gene symbols using the biomaRt package, and
the expression of each tissue/compartment was presented against a background of all
proteins detected in bulk and/or turbo samples, mirroring the pathway analysis (above).
Missing values were imputed as zeros, and ImmuCellAI was run with default settings.
The "infiltration score" was extracted from the resulting abundance matrix for plotting
via ggplot2.

Sexual dimorphism

Male/Female differences were tested using a standard limma framework for the
neuronal proteome of each tissue[51]. Proteins were modelled using a grouped factor for
Tissue and Sex, and contrasted within Tissue as M – F. Proteins were considered
enriched when FDR < 0.05, and abs(LFC > 1).

Weighted co-expression network analysis

A weighted protein co-expression network was built using the WGCNA package in R
(version 1.72-5)[54]. Turbo samples were filtered for the entire universe of enriched
proteins (ie., enriched in any component). Proteins expressed only in a single
component and proteins with a variance of 0 were then removed, resulting in a matrix of
3142 proteins across 39 samples.

Soft power was established for a signed network using `pickSoftThreshold(type =
"signed")`, with a model fit of 0.8. The scale-free topology fit index failed for reasonable
powers (< 30), suggesting a strong correlation among large gene groups, thus violating
the assumption of the scale-free topology approximation. As noted in the WGCNA
package documentation, this is commonly seen with datasets containing multiple
tissue types, and a lack of scale-free topology fit does not inherently invalidate the

data[54]. Consequently, we selected the recommended soft power of 14 for signed
networks with 30-40 samples.

Network adjacency was calculated with `adjacency(type = "signed")` and the
topological overlap matrix was calculated with `TOMsimilarity(TOMtype = "signed")`,
both using default settings. Modules were established through adaptive branch pruning
`dynamicMods <- cutreeDynamic(dendro = geneTree, distM = dissTOM, deepSplit = 2,
pamRespectsDendro = FALSE, minClusterSize = 30)`. Module eigengenes were then
extracted for downstream analyses.

Terminal protein-protein interaction networks

Neuronally enriched proteins (the "neuronal proteome") for each terminal
compartment were used to generate large-scale protein-protein interaction networks in
Cytoscape, via the stringApp[55,56]. For the peripheral terminals, a core cluster of 433
proteins were highly connected through high confidence (0.7) protein-protein
interactions. For the central terminals, 2271 proteins formed the core cluster. For each
compartment, these networks were then clustered further. Here, community clustering
was performed on the resulting STRING network in Cytoscape, where the Girvan-
Newman fast greedy algorithm was implemented through ClusterMaker2 via GLay[21,22].
Functional attributes within each cluster where then calculated via the stringApp,
against a background of the entire peripheral/central terminal proteome respectively.
Cluster numbers are somewhat arbitrary, but correspond to the supplemental table
outputs with full enrichment details.

Oxaliplatin effects

The effects of the chemotherapy agent oxaliplatin on explant cultures was tested using
limma.[51] For DRG explants dissected from wildtype mice, oxaliplatin was administered
at 25 µM, 50 µM and 100 µM concentrations (or vehicle). Proteins here were modelled as
Concentration.Condition, and contrasted within each concentration. For TurboID mice,
data were first filtered for Turbo samples and neuronally-enriched proteins prior to
modelling by Condition. In each case, limma was used to contrast by treatment
condition (oxaliplatin – vehicle). Proteins were considered differentially expressed with
FDR < 0.05, and abs(LFC) > 0.5. Downstream processing is described in the `*Pathway
analyses*` section.

Comparisons against LFC and overall expression were compared to previously
published, DIA-MS from *in vivo* oxaliplatin dosing by Yang *et al,* 2022.[27]

Rank-Rank Hypergeometric Overlap

Rank-Rank Hypergeometric Overlap (RRHO) testing was performed on limma outputs.
Here, concordant and discordant gene profiles were calculated via the Stratified
method using ranked `$-\log_{10}$(p-values) * sign(effect size)`, implemented through the
RRHO2 library in R in line with their published recommendations.[57]

STRING analysis

STRING protein-protein interaction analyses were performed through Cytoscape 3.10.3 using the stringApp.[55,56] A list of DEPs for each significant biological pathway were extracted from the ClusterProfiler output (see: *Pathway analyses*) and loaded into the stringApp with default settings. Proteins were coloured based on differential expression in either TurboID only, or TurboID + bulk explant datasets.

Immunohistochemistry (IHC)

For IHC the tissue was dissected as above with modifications: instead of the dorsal part of the SC, the whole L3-5 SC was taken out. The plantar paw punch contained not only the epidermis and dermis but all layers below the bones. The dissected tissue was fixed in 4% formalin/PBS for 3 h at 4°C, followed by overnight cryoprotection in 30% (w/v) sucrose at 4°C. For sectioning, the tissue was embedded in Tissue-Tek O.C.T. compound and frozen. Serial sections (SCN, DRG: 10 μm, SC: 30 μm, skin: 20-40 μm) were obtained using a cryostat and mounted on SuperFrost Plus slides. Sections were stored at -80°C until further processing.

Frozen sections were thawed at RT for 30 min before permeabilization and blocking in 0.4% Triton X-100/PBS containing 5% donkey serum for 30 min. Primary antibodies, diluted in 0.1% Triton X-100/PBS with 1% donkey serum, were applied overnight at 4°C. Following three washes in PBS, sections were incubated at RT for 2 h with Alexa Fluor-conjugated secondary antibodies in PBS containing 1% donkey serum and 0.1% Triton X-100. Unbound antibodies were removed by washing with PBS, and sections were mounted using SlowFade™ Glass Soft-set Antifade Mountant Medium containing 4',6-Diamidino-2-phenylindole (DAPI). Antibodies used are listed in STable 1.

Graphics

R base functions and the following libraries were used to compile plots unless otherwise described: ggplot2, ggbiplot, ComplexHeatmap, cowplot. Figures and schematics were designed in Inkscape.

Use of generative AI

Generative AI tools (OpenAI ChatGPT) were used solely to improve phrasing, clarity, and formatting. All scientific concepts, experimental designs, hypotheses, and project goals were developed by the research team. No synthetic or AI-generated data were included.

References

1. Middleton SJ, Barry AM, Comini M, et al. Studying human nociceptors: From fundamentals to clinic. *Brain*. 2021. doi:10.1093/brain/awab048

2. Qin W, Cho KF, Cavanagh PE, Ting AY. Deciphering molecular interactions by proximity labeling. *Nat Methods*. 2021;18(2). doi:10.1038/s41592-020-01010-5

3. Gingras AC, Abe KT, Raught B. Getting to know the neighborhood: using proximity-dependent biotinylation to characterize protein complexes and map organelles. *Curr Opin Chem Biol*. 2019;48. doi:10.1016/j.cbpa.2018.10.017

4. Branon TC, Bosch JA, Sanchez AD, et al. Efficient proximity labeling in living cells and organisms with TurboID. *Nat Biotechnol*. 2018;36(9). doi:10.1038/nbt.4201

5. Roux KJ, Kim DI, Raida M, Burke B. A promiscuous biotin ligase fusion protein identifies proximal and interacting proteins in mammalian cells. *J Cell Biol*. 2012;196(6). doi:10.1083/jcb.201112098

6. Dingar D, Kalkat M, Chan PK, et al. BioID identifies novel c-MYC interacting partners in cultured cells and xenograft tumors. *J Proteomics*. 2015;118. doi:10.1016/j.jprot.2014.09.029

7. May DG, Scott KL, Campos AR, Roux KJ. Comparative Application of BioID and TurboID for Protein-Proximity Biotinylation. *Cells*. 2020;9(5). doi:10.3390/cells9051070

8. Sunna S, Bowen C, Zeng H, et al. Cellular Proteomic Profiling Using Proximity Labeling by TurboID-NES in Microglial and Neuronal Cell Lines. *Mol Cell Proteomics*. 2023;22(6). doi:10.1016/j.mcpro.2023.100546

9. Rayaprolu S, Bitarafan S, Santiago J V., et al. Cell type-specific biotin labeling in vivo resolves regional neuronal and astrocyte proteomic differences in mouse brain. *Nat Commun*. 2022;13(1). doi:10.1038/s41467-022-30623-x

10. Hobson BD, Choi SJ, Mosharov E V., Soni RK, Sulzer D, Sims PA. Subcellular proteomics of dopamine neurons in the mouse brain. *Elife*. 2022;11. doi:10.7554/eLife.70921

11. Zurborg S, Piszczek A, Martínez C, et al. Generation and characterization of an Advillin-Cre driver mouse line. *Mol Pain*. 2011;7. doi:10.1186/1744-8069-7-66

12. Sisignano M, Baron R, Scholich K, Geisslinger G. Mechanism-based treatment for chemotherapy-induced peripheral neuropathic pain. *Nat Rev Neurol*. 2014;10(12). doi:10.1038/nrneurol.2014.211

13. Barry AM, Zhao N, Yang X, Bennett DL, Baskozos G. Deep RNA-seq of male and female murine sensory neuron subtypes after nerve injury. *Pain*. 2023;164(10):2196-2215. doi:10.1101/2022.11.21.516781

14. Zheng Y, Liu P, Bai L, Trimmer JS, Bean BP, Ginty DD. Deep Sequencing of Somatosensory Neurons Reveals Molecular Determinants of Intrinsic Physiological Properties. *Neuron*. 2019;103(4):598-616.e7. doi:10.1016/j.neuron.2019.05.039

15. Lopes DM, Denk F, McMahon SB. The Molecular Fingerprint of Dorsal Root and Trigeminal Ganglion Neurons. *Front Mol Neurosci*. 2017. doi:10.3389/fnmol.2017.00304

16. Sun X, Sun H, Han X, et al. Deep Single-Cell-Type Proteome Profiling of Mouse Brain by Nonsurgical AAV-Mediated Proximity Labeling. *Anal Chem*. 2022;94(13). doi:10.1021/acs.analchem.1c05212

17. Dumrongprechachan V, Salisbury RB, Soto G, Kumar M, MacDonald ML, Kozorovitskiy Y. Cell-type and subcellular compartment-specific APEX2 proximity labeling reveals activity-dependent nuclear proteome dynamics in the striatum. *Nat Commun*. 2021;12(1). doi:10.1038/s41467-021-25144-y

18. Bhuiyan SA, Xu M, Yang L, et al. Harmonized cross-species cell atlases of trigeminal and dorsal root ganglia. *Sci Adv*. 2024;10(25). doi:10.1126/sciadv.adj9173

19. Tavares-Ferreira D, Shiers S, Ray PR, et al. Spatial transcriptomics of dorsal root ganglia identifies molecular signatures of human nociceptors. *Sci Transl Med*. 2022;14(632):eabj8186. doi:DOI: 10.1126/scitranslmed.abj8186

20. Szklarczyk D, Gable AL, Nastou KC, et al. The STRING database in 2021: customizable

994 protein–protein networks, and functional characterization of user-uploaded
995 gene/measurement sets. *Nucleic Acids Res*. 2021;49(D1):D605.
996 doi:10.1093/NAR/GKAA1074

997 21. Utriainen M, Morris JH. clusterMaker2: a major update to clusterMaker, a multi-algorithm
998 clustering app for Cytoscape. *BMC Bioinformatics*. 2023;24(1). doi:10.1186/s12859-023-
999 05225-z

1000 22. Su G, Kuchinsky A, Morris JH, States DJ, Meng F. GLay: Community structure analysis of
1001 biological networks. *Bioinformatics*. 2010;26(24). doi:10.1093/bioinformatics/btq596

1002 23. Xian F, Sondermann JR, Varela DG, Schmidt M. Deep proteome profiling reveals
1003 signatures of age and sex differences in paw skin and sciatic nerve of naïve mice. *Elife*.
1004 2022;11. doi:10.7554/eLife.81431

1005 24. Barry AM, Sondermann JR, Sondermann J-H, Gomez-Varela D, Schmidt M. Region-
1006 resolved quantitative proteome profiling reveals molecular dynamics associated with
1007 chronic pain in the PNS and spinal cord. *Front Mol Neurosci*. 2018;11(August):259.
1008 doi:10.3389/fnmol.2018.00259

1009 25. Miao YR, Xia M, Luo M, Luo T, Yang M, Guo AY. ImmuCellAI-mouse: a tool for
1010 comprehensive prediction of mouse immune cell abundance and immune
1011 microenvironment depiction. *Bioinformatics*. 2022;38(3).
1012 doi:10.1093/bioinformatics/btab711

1013 26. Dyring-Andersen B, Løvendorf MB, Coscia F, et al. Spatially and cell-type resolved
1014 quantitative proteomic atlas of healthy human skin. *Nat Commun*. 2020;11(1).
1015 doi:10.1038/s41467-020-19383-8

1016 27. Yang L, Wang H, Lu W, et al. Quantitative proteomic analysis of oxaliplatin induced
1017 peripheral neurotoxicity. *J Proteomics*. 2022;266. doi:10.1016/j.jprot.2022.104682

1018 28. Matzinger M, Mayer RL, Mechtler K. Label-free single cell proteomics utilizing ultrafast LC
1019 and MS instrumentation: A valuable complementary technique to multiplexing.
1020 *Proteomics*. 2023;23(13-14). doi:10.1002/pmic.202200162

1021 29. Grindberg R V., Yee-Greenbaum JL, McConnell MJ, et al. RNA-sequencing from single
1022 nuclei. *Proc Natl Acad Sci*. 2013;110(49):19802-19807. doi:10.1073/pnas.1319700110

1023 30. Ferreira DT, Shen BQ, Mwirigi JM, et al. Deciphering the molecular landscape of human
1024 peripheral nerves: implications for diabetic peripheral neuropathy. *bioRxiv*.
1025 2024:2024.06.15.599167.
1026 https://www.biorxiv.org/content/10.1101/2024.06.15.599167v1%0Ahttps://www.biorxiv.
1027 org/content/10.1101/2024.06.15.599167v1.abstract.

1028 31. North RY, Li Y, Ray P, et al. Electrophysiological and transcriptomic correlates of
1029 neuropathic pain in human dorsal root ganglion neurons. *Brain*. 2019;142(5):1215-1226.
1030 doi:10.1093/brain/awz063

1031 32. Schulte RR, Ho RH. Organic anion transporting polypeptides: Emerging roles in cancer
1032 pharmacology. *Mol Pharmacol*. 2019;95(5). doi:10.1124/mol.118.114314

1033 33. Litke C, Hagenston AM, Kenkel AK, et al. Organic anion transporter 1 is an HDAC4-
1034 regulated mediator of nociceptive hypersensitivity in mice. *Nat Commun*. 2022;13(1).
1035 doi:10.1038/s41467-022-28357-x

1036 34. Huang KM, Leblanc AF, Uddin ME, et al. Neuronal uptake transporters contribute to
1037 oxaliplatin neurotoxicity in mice. *J Clin Invest*. 2020;130(9). doi:10.1172/JCI136796

1038 35. Filipski KK, Mathijssen RH, Mikkelsen TS, Schinkel AH, Sparreboom A. Contribution of
1039 organic cation transporter 2 (OCT2) to cisplatin-induced nephrotoxicity. *Clin Pharmacol*
1040 *Ther*. 2009;86(4). doi:10.1038/clpt.2009.139

1041 36. Sprowl JA, Ciarimboli G, Lancaster CS, et al. Oxaliplatin-induced neurotoxicity is
1042 dependent on the organic cation transporter OCT2. *Proc Natl Acad Sci U S A*.
1043 2013;110(27). doi:10.1073/pnas.1305321110

1044 37. Li Y, Drabison T, Nepal M, et al. Targeting a xenobiotic transporter to ameliorate
1045 vincristine-induced sensory neuropathy. *JCI Insight*. 2023;8(14).

doi:10.1172/jci.insight.164646

38.   Leskelä S, Jara C, Leandro-García LJ, et al. Polymorphisms in cytochromes P450 2C8 and
3A5 are associated with paclitaxel neurotoxicity. *Pharmacogenomics J*. 2011;11(2).
doi:10.1038/tpj.2010.13

39.   Schmidt D, Díaz P, Muñoz D, et al. Characterisation of the pathophysiology of neuropathy
and sensory dysfunction in a mouse model of recessive dystrophic epidermolysis
bullosa. *Pain*. 2022;163(10). doi:10.1097/j.pain.0000000000002599

40.   Cooper AH, Barry AM, Chrysostomidou P, et al. Peripheral nerve injury results in a biased
loss of sensory neuron subpopulations. *Pain*. August 2024.
doi:10.1097/j.pain.0000000000003321

41.   Perez-Riverol Y, Bai J, Bandla C, et al. The PRIDE database resources in 2022: A hub for
mass spectrometry-based proteomics evidences. *Nucleic Acids Res*. 2022;50(D1).
doi:10.1093/nar/gkab1038

42.   Tycko J, Wainberg M, Marinov GK, et al. Mitigation of off-target toxicity in CRISPR-Cas9
screens for essential non-coding elements. *Nat Commun*. 2019;10(1).
doi:10.1038/s41467-019-11955-7

43.   Doench JG, Fusi N, Sullender M, et al. Optimized sgRNA design to maximize activity and
minimize off-target effects of CRISPR-Cas9. *Nat Biotechnol*. 2016;34(2).
doi:10.1038/nbt.3437

44.   Madisen L, Zwingman TA, Sunkin SM, et al. A robust and high-throughput Cre reporting
and characterization system for the whole mouse brain. *Nat Neurosci*. 2010;13(1):133-
140. doi:10.1038/NN.2467

45.   Concordet JP, Haeussler M. CRISPOR: Intuitive guide selection for CRISPR/Cas9 genome
editing experiments and screens. *Nucleic Acids Res*. 2018;46(W1).
doi:10.1093/nar/gky354

46.   Haeussler M, Schönig K, Eckert H, et al. Evaluation of off-target and on-target scoring
algorithms and integration into the guide RNA selection tool CRISPOR. *Genome Biol*.
2016;17(1). doi:10.1186/s13059-016-1012-2

47.   Livni L, Lees JG, Barkl-Luke ME, Goldstein D, Moalem-Taylor G. Dorsal root ganglion
explants derived from chemotherapy-treated mice have reduced neurite outgrowth in
culture. *Neurosci Lett*. 2019;694. doi:10.1016/j.neulet.2018.11.016

48.   Hughes CS, Moggridge S, Müller T, Sorensen PH, Morin GB, Krijgsveld J. Single-pot, solid-
phase-enhanced sample preparation for proteomics experiments. *Nat Protoc*.
2019;14(1). doi:10.1038/s41596-018-0082-x

49.   Demichev V, Messner CB, Vernardis SI, Lilley KS, Ralser M. DIA-NN: neural networks and
interference correction enable deep proteome coverage in high throughput. *Nat
Methods*. 2020;17(1). doi:10.1038/s41592-019-0638-x

50.   Cox J, Hein MY, Luber CA, Paron I, Nagaraj N, Mann M. Accurate proteome-wide label-
free quantification by delayed normalization and maximal peptide ratio extraction,
termed MaxLFQ. *Mol Cell Proteomics*. 2014;13(9). doi:10.1074/mcp.M113.031591

51.   Ritchie ME, Phipson B, Wu D, et al. Limma powers differential expression analyses for
RNA-sequencing and microarray studies. *Nucleic Acids Res*. 2015;43(7).
doi:10.1093/nar/gkv007

52.   Wu T, Hu E, Xu S, et al. clusterProfiler 4.0: A universal enrichment tool for interpreting
omics data. *Innov*. 2021;2(3):100141. doi:10.1016/J.XINN.2021.100141

53.   Love MI, Huber W, Anders S. Moderated estimation of fold change and dispersion for
RNA-seq data with DESeq2. *Genome Biol 2014 1512*. 2014;15(12):1-21.
doi:10.1186/S13059-014-0550-8

54.   Langfelder P, Horvath S. WGCNA: An R package for weighted correlation network
analysis. *BMC Bioinformatics*. 2008;9. doi:10.1186/1471-2105-9-559

55.   Doncheva NT, Morris JH, Holze H, et al. Cytoscape stringApp 2.0: Analysis and
Visualization of Heterogeneous Biological Networks. *J Proteome Res*. 2023;22(2).

1098   doi:10.1021/acs.jproteome.2c00651

1099   56.   Shannon P, Markiel A, Ozier O, et al. Cytoscape: A software Environment for integrated

1100   models of biomolecular interaction networks. *Genome Res*. 2003;13(11).

1101   doi:10.1101/gr.1239303

1102   57.   Cahill KM, Huo Z, Tseng GC, Logan RW, Seney ML. Improved identification of concordant

1103   and discordant gene expression signatures using an updated rank-rank hypergeometric

1104   overlap approach. *Sci Rep*. 2018;8(1). doi:10.1038/s41598-018-27903-2

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1108 **Supplemental Tables + Figures: Proximity labelling reveals the compartmental**
1109 **proteome of murine sensory neurons**
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Order	Content
Supplemental Table 1	expression matrix, TurboID (compartments)
Supplemental Table 2	neuronal enrichment
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Supplemental Table 5	GO compartment functions
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Supplemental Table 7	PPI peripheral terminal functions
Supplemental Table 8	PPI central terminal functions
Supplemental Table 9	expression matrix, bulk (tissues)
Supplemental Table 10	TurboID unique proteins
Supplemental Table 11	expression matrix, explant (bulk)
Supplemental Table 12	DEPs (bulk oxaliplatin)
Supplemental Table 13	expression matrix, explant (TurboID)
Supplemental Table 14	neuronal enrichment, explant
Supplemental Table 15	DEPs (TurboID oxaliplatin)
Supplemental Table 16	GO oxaliplatin treated-explant functions
Supplemental Table 17	Key resources

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1115 **Supplemental Table 17**. Key Resources

Type	Designation	Source	Identifiers	Additional
Mouse strain	C57BL/6J	In house bred		Wild type, male
Mouse strain	Advillin^{Cre}	Kind gift of Prof. Dr. Klaus-Armin Nave, MPIEM, Göttingen	(Zurborg et al., 2011) Tg(Avil-cre)1Phep MGI:5292346	
Mouse strain	TurboID$^{fl/fl}$	Current study	ROSA26$^{em1(TurboID)Bros}$	
Chemical	Acetonitrile	Fisher Scientific	10001334	
Chemical	Formic Acid	Fisher Scientific	15658430	
Chemical	10x PBS	Fisher Scientific	11594516	
Chemical	Tris 1M	Accugene/Avantor	51237	
Chemical	NaCl, 5M	Accugene/Avantor	51202	
Chemical	Triton™ X-100	Fisher Scientific	10102913	

Chemical	Invitrogen™ SDS, 20% Solution, RNase-free	Fisher Scientific	AM9820	
Chemical	Biotin	Sigma-Aldrich	B4501-10G	
Chemical	Oxaliplatin	Sigma-Aldrich	O9512-5MG	
Chemical	Glycerol	Fisher Scientific	10021083	
Chemical	Dithiothreitol 1M	Sigma-Aldrich	43816	
Chemical	Acetone	Sigma-Aldrich	1000201000	
Chemical	Ethanol	Sigma-Aldrich	1117272500	
Chemical	Iodoacetamide	Sigma-Aldrich	I1149	
Chemical	Ammonium bicarbonate	Sigma-Aldrich	09830-500G	
Chemical	Water MS grade	Sigma-Aldrich	1.15333.1000	
other	Trypsin/Lys-C	Promega	V5073	
other	rhβ-Nerve Growth Factor	R&D Systems	256-GF	
other	Gibco™ DMEM/F-12, GlutaMAX™ Supplement	Fisher Scientific	31331028	
other	Sera-Mag SpeedBead beads	Cytiva	65152105050250, 45152105050250	1:1 mix
other	Sera-Mag SpeedBead Blocked Streptavidin Particles	Cytiva	21152104010150	
other	Complete protease inhibitor cocktail	Roche/ Merck	58929700001	Mix of protease inhibitors
other	Protein LoBind tube	Eppendorf	0030108116, 0030108132	Reagent tube
other	4X Bolt™ LDS sample buffer	Fisher Scientific	B0007	
other	10X Bolt™ Sample Reducing Agent	Fisher Scientific	B0009	
other	Bolt™ Bis-Tris Plus Mini Protein Gels, 4-12%, 1.0 mm	Fisher Scientific	NW04120BOX	
other	Invitrogen™ iBlot™ 2 Transfer Stacks, PVDF, mini	Fisher Scientific	IB24002	
other	SlowFade™ Glass Soft-set Antifade Mountant, with DAPI	Fisher Scientific	S36920	

other	Aurora Series UHPLC column	IonOpticks	AUR2-25075C18A-CSI	
other	Biopsy punch 4mm	Kai medical	48401	
antibody	Streptavidin-AF680	invitrogen	S21378	1:500 dilution
antibody	Mouse mc α-β-actin	Sigma-Aldrich	A1978-200UL	1:500 dilution
antibody	Streptavidin-AF555	Invitrogen	S32355	1:500 dilution
antibody	Rabbit α-NF200	Sigma-Aldrich	N4142-25UL	1:500 dilution
antibody	Chicken α-Peripherin	ABcam	AB39374	1:100 dilution
antibody	Donkey α-mouse AF790	Invitrogen	A11371	1:8000 dilution
antibody	Donkey α-rabbit 488	Invitrogen	A-21206	1:250 dilution
antibody	Donkey α-chicken 647	Invitrogen	A78952	1:250 dilution
software, algorithm	DIA-NN	https://github.com/vdemichev/DiaNN		1.8.2
software, algorithm	R	https://www.r-project.org/		4.3.2
software, algorithm	Mouse proteome	Uniprot	UP000000589	
software, algorithm	Cytoscape	https://cytoscape.org/		3.10.3

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1118 **Supplemental Figures**

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1122 **Supplemental Figure 1.** A. Immunohistochemistry of Adv[Cre];TurboID[fl/fl] with classic sensory

1123 neuron markers peripherin and NF200. Scale bar = 50 um. B. Western Blot analyses for TurboID

1124 (left) and B-actin loading controls (right). Adv[Cre];TurboID[fl/fl] (T) and TurboID[fl/fl] (TC) samples for

1125 each tissue were loaded. DRG, dorsal root ganglia; SCN, sciatic nerve; SC, lumbar spinal cord.

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Supplemental Figure 2. DIA-PASEF mass spectrometry of TurboID and control samples across
sensory neuron compartments. A-B. Summed expression (A), and Protein Group counts (B) by
tissue and TurboID status. C. PCA across all samples. D. Heatmap of ion channel expression
across all samples. E. Dynamic range for proteins of interest in the Adv[Cre];TurboID[fl/fl], plotted as
expression (log$_2$ Intensity) per component. terminals$_p$ = peripheral terminals (skin); terminals$_c$ =
central terminals (SC). Relates to Figure 2.

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Supplemental Figure 3. Sexual dimorphism across neuronal compartments. A. Differential
protein expression by component, comparing male vs female mice within TurboID samples
enriched for the neuronal proteome. Significant proteins are in pink. LFC = $\log_2$ fold change B.
Biological Pathway analysis for the paw and DRG on ranked fold changes between male and
female samples. Counts reflect the number of proteins in each respective term. terminals$_p$ =
peripheral terminals; terminals$_c$ = central terminals.
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1150 **Supplemental Figure 4.** Community clustering of the central terminal proteome, with edges

1151 denoting high confidence protein-protein interactions. Protein nodes are coloured by cluster,

1152 and clusters are described by functional summaries with proteins of interest highlighted. Full

1153 functional enrichments are available at Stable 8.

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1157 **Supplemental Figure 5.** DIA-PASEF tissue proteomes along the peripheral somatosensory axis.

1158 A. Protein Group (PG) counts per tissue. B. PCA by tissue. C. Heatmap of key neuronal genes. D.

1159 Comparison between TurboID and wildtype, whole tissue (WT) proteomes for key neuronal

1160 proteins.

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1163 **Supplemental Figure 6.** Depletion of ImmuCellAI immune cell signatures between bulk (WT)
1164 and turbo (Adv[Cre];Turbo[fl/fl]) samples for each tissue/compartment.

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Supplemental Figure 7. DIA-PASEF proteomics of DRG explants exposed to Oxaliplatin (25, 50
or 100 uM) or Vehicle (Veh). A. Protein Group counts. B. PCA across sample conditions. C.
Correlation across samples, labelled as Oxaliplatin (25, 50, 100) or Vehicle (Veh). D. Heatmap
across key neuronal proteins. E. Differentially expressed proteins between 100 uM Oxaliplatin
and Vehicle-treated DRG explants. Purple dots are significant (FDR < 0.05, LFC > 0.5). F.
Correlation of protein expression with previously published, *in vivo* treated DRGs (Yang et al.,
2022). Rank-rank hypergeometric overlap between oxaliplatin changes across datasets (Yang et
al., 2022 and here). DEP, differentially expressed proteins; LFC, $\log_2$ fold change.
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1178 **Supplemental Figure 8.** DIA-PASEF proteomics of DRG explants from Adv^Cre^;TurboID^fl/fl^ and
1179 TurboID^fl/fl^ mice. A. Protein group counts across conditions, before neuronal enrichment. B. PCA
1180 across explant samples. C. Differentially expressed proteins between Adv^Cre^;TurboID^fl/fl^ and
1181 TurboID^fl/fl^ explant samples, for both vehicle and oxaliplatin-treated cultures. D. Neuronal
1182 enrichments for each condition. LFC = log$_2$ fold change. Ox = oxaliplatin. V = vehicle.
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Supplemental Methods: Proximity labelling reveals the compartmental proteome of murine sensory neurons

Generation of TurboID mutant mice using CRISPR/Cas9

The conditional TurboID knock-in (cKI) mouse line was generated by site-directed CRISPR/Cas9 mutagenesis in the ROSA26 locus. Superovulated C57Bl6/J females were mated with C57Bl6/J males and fertilized eggs collected. In-house produced CRISPR reagents (hCas9_mRNA, sgRNAs, preformed Cas9_sgRNA-RNP complexes and the homology-directed repair fragment (HDR fragment)) were microinjected into the pronucleus and the cytoplasm of pronuclear-stage zygotes using an Eppendorf Femtojet. Importantly, all nucleotide-based CRISPR/Cas9 reagents (sgRNAs and hCAS9_mRNA) were used as RNA molecules and were not plasmid-encoded, reducing the probability of off-target effects due to the short lifetime of RNA-based reagents (Tycko et al., 2019; Doench et al., 2016). The TurboID-containing HDR fragment was generated by subcloning a V5-tag-TurboID sequence into the Ai14-ROSA26-based plasmid backbone (Madisen et al., 2010) and isolated as a linear DNA fragment by SacI/AgeI restriction enzyme double digestion. Two sgRNAs targeting the ROSA26 Locus were selected using the guide RNA selection tool CRISPOR (Haeussler et al, 2016; Concordet & Haeussler, 2018). The correct locus-specific insertion of the HDR fragment was confirmed by localization PCRs with primers upstream and downstream of the HDR sequence.

Genotyping Strategy

For genotyping, genomic DNA (gDNA) was isolated from tail biopsies using a genomic DNA isolation kit (Nexttec, #10.924). For the location PCRs, 20 µL reactions were prepared using 1 µL clean gDNA (15-80ng), 4 µL PrimerSet (4 pmol final each), 4 µL 5X Reaction Buffer (Finnzymes #F-524), 0.4 µL PhireHot-Start II Taq DNA Polymerase (Finnzymes #F-122L), 1 µL 10 mM dNTPs (Bioline #DM-515107), 4 µL Hi-Spec Additive (Bioline #HS-014101) and 5.6 µL H2O. Thermocycler parameters location-PCR Rosa26 left arm: 98°C for 5 min, (98 °C for 45 s, 64 °C for 30 s, 72 °C for 60 s) repeated for 34 cycles, final step at 72 °C for 10 min. Thermocycler parameters-location PCR Rosa26 right arm: 98°C for 5 min, (98 °C for 45 s, 68 °C for 30 s, 72 °C for 180 s) repeated for 34 cycles, final step at 72 °C for 10 min.

For the diagnostic routine genotyping PCR, 20 µL reactions were prepared using 1 µL clean gDNA (15-80ng), 4 µL PrimerSet (4 pmol final each), 4 µL 5X Reaction Buffer (Biozym #331620XL), 0.2 µL Hot-Start Taq DNA Polymerase (Biozym #331620XL), 1 µL 50 mM MgCl2 (AGCTLab stock) and 9.8 µL H2O. Thermocycler parameters: 96°C for 3 min, (94 °C for 30 s, 62 °C for 60 s, 72 °C for 60 s) repeated for 32 cycles, final step at 72 °C for 7 min.

Rosa26 Locus sgRNAs
Sense Rosa26 protospacer-sgRNA1 sequence: 5'-ACTCCAGTCTTTCTAGAAGA-3' (PAM = TGG)
Antisense Rosa26 protospacer-sgRNA2 sequence: 5'-CGCCCATCTTCTAGAAAGAC-3' (PAM = TGG)

1238   **C57BL6J_Chrom6_Rosa26 target locus (insertion site):**

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1242   **Ai14-TurboID_Rosa26 HDR fragment targeting strategy (conditional knock-in)**

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1246   **Ai14-TurboID_Rosa26 Cre recombined (knock-in)**

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1249   **SMethods, Figure1. Schematic representation of the target locus, the targeting strategy**
1250   **(HDR fragment) and the Cre recombination event.**

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1252   *Ai14-TurboID_Rosa26 HDR fragment and sequence annotations:*
1253   Lower case letters = upstream and downstream HDR fragment ROSA26 sequences
1254   Capital letters = HDR fragment after locus-specific insertion

Magenta = ROSA26 left- and right arm sequence
Blue = CAG-promoter
Red BOLD = loxP sites
<u>Underlined BOLD</u> = loxP flanked STOP cassette
Green BOLD = ATG Start- and TAA Stop codon
Brown BOLD = V5-tag
Purple BOLD = TurboID
Black BOLD = WPRE
<u>Black underlined</u> = bGH poly(A) signal

1264

5'-

ggaggctgccggggccgcctaaagaagaggctgtgctttggggctccggctcctcagagagcctcggctaggtaggggatcgggactctggcggg

agggcggcttggtgcgtttgcggggatgggcggccGCGGCAGGCCCTCCGAGCGTGGTGGAGCCGTTCTGTGAGACAGC

CGGGTACGAGTCGTGACGCTGGAAGGGGCAAGCGGGTGGTGGGCAGGAATGCGGTCCGCCCTGCAGCAA

CCGGAGGGGGAGGGAGAAGGGAGCGGAAAAGTCTCCACCGGACGCGGCCATGGCTCGGGGGGGGGGG

GGCAGCGGAGGAGCGCTTCCGGCCGACGTCTCGTCGCTGATTGGCTTCTTTTCCTCCCGCCGTGTGTGAAA

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CAGCCTCGCTCTGCCCACTGGGTGGGGCGGGAGGTAGGTGGGGTGAGGCGAGCTGGACGTGCGGGCGC

GGTCGGCCTCTGGCGGGGCGGGGGAGGGGAGGGAGGGTCAGCGAAAGTAGCTCGCGCGCGAGCGGCC

GCCCACCCTCCCCTTCCTCTGGGGGAGTCGTTTTACCCGCCGCCGGCCGGGCCTCGTCGTCTGATTGGCT

CTCGGGGCCCAGAAAACTGGCCCTTGCCATTGGCTCGTGTTCGTGCAAGTTGAGTCCATCCGCCGGCCAGC

GGGGGCGGCGAGGAGGCGCTCCCAGGTTCCGGCCCTCCCCTCGGCCCCGCGCCGCAGAGTCTGGCCG

CGCGCCCCTGCGCAACGTGGCAGGAAGCGCGCGCTGGGGGCGGGGACGGGCAGTAGGGCTGAGCGGC

TGCGGGGCGGGTGCAAGCACGTTTCCGACTTGAGTTGCCTCAAGAGGGGCGTGCTGAGCCAGACCTCCATC

GCGCACTCCGGGGAGTGGAGGGAAGGAGCGAGGGCTCAGTTGGGCTGTTTTGGAGGCAGGAAGCACTTGC

TCTCCCAAAGTCGCTCTGAGTTGTTATCAGTAAGGGAGCTGCAGTGGAGTAGGCGGGGAGAAGGCCGCACC

CTTCTCCGGAGGGGGGAGGGGAGTGTTGCAATACCTTTCTGGGAGTTCTCTGCTGCCTCCTGGCTTCTGAGG

ACCGCCCTGGGCCTGGGAGAATCCCTTCCCCCTCTTCCCTCGTGATCTGCAACTCCAGTCTTTCTAGCCTTAA

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TATGGAGTTCCGCGTTACATAACTTACGGTAAATGGCCCGCCTGGCTGACCGCCCAACGACCCCCGCCCATT

GACGTCAATAATGACGTATGTTCCCATAGTAACGCCAATAGGGACTTTCCATTGACGTCAATGGGTGGAGTATTT

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GGTAAATGGCCCGCCTGGCATTATGCCCAGTACATGACCTTATGGGACTTTCCTACTTGGCAGTACATCTACG

TATTAGTCATCGCTATTACCATGGTCGAGGTGAGCCCCACGTTCTGCTTCACTCTCCCCATCTCCCCCCCCTC

CCCACCCCCAATTTTGTATTTATTTATTTTTTAATTATTTTGTGCAGCGATGGGGGCGGGGGGGGGGGGGGGGGC

GCGCGCCAGGCGGGGCGGGGCGGGGCGAGGGCGGGGCGGGGCGAGGCGGAAAGGTGCGGCGGCA

GCCAATCARAGCGGCGCGCTCCGAAAGTTTCCTTTTATGGCGAGGCGGCGGCGGCGGCGGCCCTATAAAAA

GCGAAGCGCGCGGCGGGCGGGAGTCGCTGCGCGCTGCCTTCGCCCCGTGCCCCGYTCCGCCGCCGCCT

CGCGCCGCCCGCCCCGGYTCTGACTGACCGCGTTACTCCCACAGGTGAGCGGGCGGGACGGCCCTTCTC

CTCCGGGYTGTAATTAGCGCTTGGTTTAATGACGGYTTGTTTCTTTTCTGTGGYTGCGTGAAAGCCTTGAGGGGC

TCCGGGAGGGCCCTTTGTGCGGGGGGAGCGGCTCGGGGGGTGCGTGCGTGTGTGTGCGTGGGGAGCG

CCGCGTGCGGCTCCGCGCTGCCCGGCGGCTGTGAGCGCTGCGGGCGCGGCGCGGGGCTTTGTGCGCTC

CGCAGTGTGCGCGAGGGGAGCGCGGCCGGGGGCGGTGCCCCGCGGTGCGGGGGGGGCTGCGAGGGG

AACAAAGGCTGCGTGCGGGGTGTGTGCGTGGGGGGGTGAGCAGGGGGTGTGGGCGCGTCGGTCGGGCTG

CAACCCCCCCTGCACCCCCCTCCCCGAGTTGCTGAGCACGGCCCGGCTTCGGGTGCGGGGCTCCGTACG

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CCAAATCTGTGCGGAGCCGAAATCTGGGAGGCGCCGCCGCACCCCCTCTAGCGGGCGCGGGGCGAAGC

GGTGCGGCGCCGGCAGGAAGGAAATGGGCGGGGAGGGCCTTCGTGCGTCGCCGCGCCGCCGTCCCCTT

CTCCCTCTCCAGCCTCGGGGCTGTCCGCGGGGGACGGCTGCCTTCGGGGGGGACGGGGCAGGGCGGG

GTTCGGCTTCTGGCGTGTGACCGGCGGCTCTAGAGCCTCTGCTAACCATGTTCATGCCTTCTTCTTTTTCCTAC

AGCTCCTGGGCAACGTGCTGGTTATTGTGCTGTCTCATCATTTTGGCAAAGAATTGATTTGATACCGCGGGCCC

TAAGAAGTTCCTATTCTCTAGAAAGTATAGGAACTTCGTCGACATTTAAATCATTTAAAT**ATAACTTCGTATAATGT**

ATGCTATACGAAGTTATTCGCGATGAATAAATGAAAGCTTGCAGATCTGCGACTCTAGAGGATCTGCGACTC

TAGAGGATCATAATCAGCCATACCACATTTGTAGAGGTTTTACTTGCTTTAAAAAACCTCCCACACCTCCCC

CTGAACCTGAAACATAAAATGAATGCAATTGTTGTTGTTAACTTGTTTATTGCAGCTTATAATGGTTACAAATA

AAGCAATAGCATCACAAATTTCACAAATAAAGCATTTTTTTCACTGCATTCTAGTTGTGGTTTGTCCAAACTCA

TCAATGTATCTTATCATGTCTGGATCTGCGACTCTAGAGGATCATAATCAGCCATACCACATTTGTAGAGGTT

TTACTTGCTTTAAAAAACCTCCCACACCTCCCCCTGAACCTGAAACATAAAATGAATGCAATTGTTGTTGTTA

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AGCATCACAAATTTCACAAATAAAGCATTTTTTTCACTGCATTCTAGTTGTGGTTTGTCCAAACTCATCAATGT

ATCTTATCATGTCTGGATCCCCATCAAGCTGATCCGGAACCCTTAAT**ATAACTTCGTATAATGTATGCTATAC**

GAAGTTATTAGGTCCCTCGACCTGCAGCCCAAGCTAGATCGAATTCGGCCGGCCTTGTACGCGTGCCACC**A**

TG**GGCAAGCCCATCCCCAACCCCCTGCTGGGCCTGGACAGCACC**GCTAGC**AAAGACAATACTGTGCCT**

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TTTTCGGGATTAGCCGGGGGATTGATAAACAGGGAGCCCTGCTGCTGGAACAGGACGGAGTTATCAAACC

CTGGATGGGCGGAGAAATCAGTCTGCGGTCTGCCGAAAAG**TAA**ACGCGTACAAGGCCGGCCCTGCAGGA

ATTCGATATCAAGCTTATCGAT**AATCAACCTCTGGATTACAAAATTTGTGAAAGATTGACTGGTATTCTTAACTA**

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TCATTTTCTCCTCCTTGTATAAATCCTGGTTGCTGTCTCTTTATGAGGAGTTGTGGCCCGTTGTCAGGCAACG

TGGCGTGGTGTGCACTGTGTTTGCTGACGCAACCCCCACTGGTTGGGGCATTGCCACCACCTGTCAGCTC

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AATCCAGCGGACCTTCCTTCCCGCGGCCTGCTGCCGGCTCTGCGGCCTCTTCCGCGTCTTCGCCTTCGC

CCTCAGACGAGTCGGATCTCCCTTTGGGCCGCCTCCCCGCATCGATACCGTCGACCTCGACCTCGA<u>CTGT</u>

<u>GCCTTCTAGTTGCCAGCCATCTGTTGTTTGCCCCTCCCCCGTGCCTTCCTTGACCCTGGAAGGTGCCACTCC</u>

<u>CACTGTCCTTTCCTAATAAAATGAGGAAATTGCATCGCATTGTCTGAGTAGGTGTCATTCTATTCTGGGGGGTGG</u>

<u>GGTGGGGCAGGACAGCAAGGGGGAGGATTGGGAAGACAATGGCAGGCATGCTGGGGA</u>ACTAGTGGTGCCA

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CATATATCAAGGCAAAACATGTTATATATGTTAAACATTTGTACTTAATGTGAAAGTTAGGTCTTTGTGGGTTTGATT

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tattaattgtgtaattattgttttttcctcctttagatcattccttgaggacaggacagtgcttgtttaaggctatatttctgctgtctgagcagcaacaggtc

ttcgagatcaacatgatgttcat-3'

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1412 **SMethods, Sequence.** Ai14-TurboID_Rosa26 HDR fragment and sequence annotations.

1413 **SMethods, Table. Primer sequence and fragment size information**

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Name	ID	Sequence (5' → 3')	mer	Orientation	Target Region	Fragment Size
Rosa26 Left Arm	29194	CGCCTAAAGAAGAGGCTGTG	20	Sense	Upstream ROSA26 left-arm	1295 bp
	7604	CGCGGAACTCCATATATGGGCT	22	Anti-sense	5'-End-CAG-promoter	
Rosa26 Right Arm	2736	GCAGGACAGCAAGGGGGAGGAT	22	Sense	bGH poly(A) signal	4736 bp
	36502	GACCTGTTGCTGCTCAGACAGC	22	Anti-sense	Downstream Rosa26 right arm	
Diagnostic Routine Genotyping PCR	14025	CTCTGCTGCCTCCTGGCTTCT	21	Sense	ROSA26 upstream insertion site	146 bp
	24229	CGCCCACACACCAGGTTAG	19	Anti-sense	ROSA26 downstream insertion site	
	30545	AAGCTGATCCGGAACCCTTAA	21	Sense	Upstream loxP2	254 bp (WT) 370 bp (Cre)
	26512	CTACAGCTCCTGGGCAACG	19	Sense	upstream loxP1 (3'-End CAG promoter)	
	38281	CCAGGGTTTCTCCCAGCTGT	20	Anti-sense	TurboID	

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