

Genetic Disruption of WASHC4 Drives Endo-lysosomal Dysfunction and Cognitive-Movement Impairments in Mice and Humans

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SUMMARY

Mutation of the WASH complex subunit, SWIP, is implicated in human intellectual disability, but the cellular etiology of this association is unknown. We identify the neuronal WASH complex proteome, revealing a network of endosomal proteins. To uncover how dysfunction of endosomal SWIP leads to disease, we generate a mouse model of the human *WASHC4*^{c.3056C>G} mutation. Quantitative spatial proteomics of SWIP^{P1019R} mouse brain confirms that this mutation destabilizes the WASH complex and reveals significant perturbations in both endosomal and lysosomal pathways. Cellular and histological analyses confirm that SWIP^{P1019R} results in endo-lysosomal disruption and uncover indicators of neurodegeneration. We find that SWIP^{P1019R} not only impacts cognition, but also causes significant progressive motor deficits in mice. Remarkably, a retrospective analysis of SWIP^{P1019R} patients confirms this deficit in humans. Combined, these findings support the model that WASH complex destabilization, resulting from SWIP^{P1019R}, drives cognitive and motor impairments via endo-lysosomal dysfunction in the brain.

KEYWORDS: WASH complex, SWIP, endosomal trafficking, endo-lysosomal, proteomics, lipofuscin, neurodegeneration, intellectual disability, progressive motor dysfunction

INTRODUCTION

Neurons maintain precise control of their subcellular proteome using a sophisticated network of vesicular trafficking pathways that shuttle cargo throughout their elaborate processes. Endosomes function as a central hub in this vesicular relay system by coordinating protein sorting between multiple cellular compartments, including surface receptor endocytosis and recycling, as well as degradative shunting to the lysosome. How endosomal trafficking is modulated in neurons remains a vital area of research due to the unique degree of spatial segregation between organelles in neurons, and its strong implication in neurodevelopmental and neurodegenerative diseases.

In non-neuronal cells, an evolutionarily conserved complex, the Wiskott-Aldrich Syndrome protein and SCAR Homology (WASH) complex, coordinates endosomal trafficking (Derivery and Gautreau, 2010; Linardopoulou et al., 2007). WASH is composed of five core protein components: WASHC1 (aka WASH1), WASHC2 (aka FAM21), WASHC3 (aka CCDC53), WASHC4 (aka SWIP), and WASHC5 (aka Strumpellin) (encoded by genes *Washc1-Washc5*, respectively), which are broadly expressed in multiple organ systems (Alekhina et al., 2017; Kustermann et al., 2018; McNally et al., 2017; Simonetti and Cullen, 2019; Thul et al., 2017). The WASH complex plays a central role in non-neuronal endosomal trafficking by activating Arp2/3-dependent actin branching at the outer surface of endosomes to influence cargo sorting and vesicular scission (Gomez and Billadeau, 2009; Lee et al., 2016; Phillips-Krawczak et al., 2015; Piotrowski et al., 2013; Simonetti and Cullen, 2019). It also interacts with at least three main cargo adaptor complexes — the Retromer, Retriever, and COMMD/CCDC22/CCDC93 (CCC) complexes — all of which associate with distinct

65 sorting nexins to select specific cargo and enable their trafficking to other cellular
66 locations (Binda et al., 2019; Farfán et al., 2013; McNally et al., 2017; Phillips-Krawczak
67 et al., 2015; Seaman and Freeman, 2014; Singla et al., 2019). Loss of the WASH complex
68 in non-neuronal cells has detrimental effects on endosomal structure and function, as its
69 loss results in aberrant endosomal tubule elongation and cargo mislocalization (Bartuzi
70 et al., 2016; Derivery et al., 2009; Gomez and Billadeau, 2009; Gomez et al., 2012;
71 Phillips-Krawczak et al., 2015; Piotrowski et al., 2013).

72 Consistent with the association between the endosomal trafficking system and
73 pathology, dominant missense mutations in *WASHC5* (protein: Strumpellin) are
74 associated with hereditary spastic paraplegia (SPG8) (De Bot et al., 2013; Valdmanis et
75 al., 2007), and autosomal recessive point mutations in *WASHC4* (protein: SWIP) and
76 *WASHC5* are associated with syndromic and non-syndromic intellectual disabilities
77 (Assoum et al., 2020; Elliott et al., 2013; Ropers et al., 2011). In particular, an autosomal
78 recessive mutation in *WASHC4* (c.3056C>G; p.Pro1019Arg) was identified in a cohort of
79 children with non-syndromic intellectual disability (Ropers et al., 2011). Cell lines derived
80 from these patients exhibited decreased abundance of WASH proteins, leading the
81 authors to hypothesize that the observed cognitive deficits in SWIP^{P1019R} patients resulted
82 from disruption of neuronal WASH signaling (Ropers et al., 2011). However, whether this
83 mutation leads to perturbations in neuronal endosomal integrity, or how this might lead to
84 cellular changes associated with disease, are unknown.

85 Here we report the analysis of neuronal WASH and its molecular role in disease
86 pathogenesis. We use *in vivo* proximity proteomics (iBiLD) to uncover the neuronal
87 WASH proteome and demonstrate that it is highly enriched for components of endosomal

trafficking. We then generate a mouse model of the human *WASHC4*^{c.3056c>g} mutation (SWIP^{P1019R}) (Ropers et al., 2011) to discover how this mutation may alter neuronal trafficking pathways and test whether it leads to phenotypes congruent with human patients. Using an adapted, spatial proteomics approach, (Geladaki et al., 2019) coupled with a systems-level analysis of protein covariation networks, we find strong evidence for substantial disruption of neuronal endosomal and lysosomal pathways *in vivo*. Cellular analyses confirm a significant impact on neuronal endo-lysosomal trafficking *in vitro* and *in vivo*, with evidence of lipofuscin accumulation and progressive apoptosis activation, molecular phenotypes that are indicative of neurodegenerative pathology. Behavioral analyses of SWIP^{P1019R} mice at adolescence and adulthood confirm a role of WASH in cognitive processes, and reveal profound, progressive motor dysfunction. Importantly, retrospective examination of SWIP^{P1019R} patient data confirms motor dysfunction coincident with cognitive impairments in humans. Our results establish that impaired WASH complex function leads to altered neuronal endo-lysosomal function, which manifests behaviorally as cognitive and movement impairments.

RESULTS

Identification of the WASH complex proteome *in vivo* confirms a neuronal role in endosomal trafficking. While multiple mutations within the WASH complex have been identified in humans (Assoum et al., 2020; Elliott et al., 2013; Ropers et al., 2011; Valdmanis et al., 2007), how these mutations lead to neurological dysfunction remains unknown (Figure 1A). Given that previous work in non-neuronal cultured cells and non-mammalian organisms have established that the WASH complex functions in endosomal

trafficking, we first aimed to confirm whether this role was conserved in the mouse nervous system (Alekhina et al., 2017; Billadeau et al., 2010; Derivery et al., 2009; Gomez and Billadeau, 2009; Gomez et al., 2012). To discover the likely molecular functions of the neuronal WASH complex, we utilized an *in vivo* BioID (iBioID) paradigm developed in our laboratory to identify the WASH complex proteome from brain tissue (Uezu et al., 2016). BioID probes were generated by fusing a component of the WASH complex, WASH1 (gene *Washc1*), with the promiscuous biotin ligase, BioID2 (WASH1-BioID2, Figure 1B), or by expressing BioID2 alone (negative control, solubleBioID2) under the neuron-specific, human Synapsin-1 promoter. We injected adenoviruses (AAV) expressing these constructs into the cortex of wild-type postnatal day zero (P0) mice (Figure 1B). Two weeks post-injection, we administered daily subcutaneous biotin for seven days to biotinylate *in vivo* substrates. The viruses displayed efficient expression and activity in brain tissue, as evidenced by colocalization of the WASH1-BioID2 viral epitope (HA) and biotinylated proteins (Streptavidin) (Figures 1C-F). For label-free quantitative high-mass accuracy LC-MS/MS analyses, whole brain samples were collected at P22, snap-frozen, and processed as previously described (Uezu et al., 2016). A total of 2,311 proteins were identified across all three experimental replicates, which were further analyzed for those with significant enrichment in WASH1-BioID2 samples over solubleBioID2 negative controls (Table S1).

The resulting neuronal WASH proteome included 174 proteins that were significantly enriched (Fold-change ≥ 3.0 , Benjamini-Hochberg P-Adjust < 0.1 , Figure 1G). Of these proteins, we identified all five WASH complex components (Figure 1H), as well as 13 previously reported WASH complex interactors (Figure 1I) (McNally et al.,

2017; Phillips-Krawczak et al., 2015; Simonetti and Cullen, 2019; Singla et al., 2019), which provided strong validity for our proteomic approach and analyses. Additional bioinformatic analyses of the neuronal WASH proteome identified a network of proteins implicated in vesicular trafficking, including 23 proteins enriched for endosomal functions (Figure 1J) and 24 proteins enriched for endocytic functions (Figure 1K). Among these endosomal and endocytic proteins were components of the recently identified endosomal sorting complexes, CCC (CCDC93 and COMMD9) and Retriever (VPS35L) (Phillips-Krawczak et al., 2015; Singla et al., 2019), as well as multiple sorting nexins important for recruitment of trafficking regulators to the endosome and cargo selection, such as SNX1-3, and SNX16 (Kvainickas et al., 2017; Maruzs et al., 2015; Simonetti et al., 2017). These data demonstrated that the WASH complex interacts with many of the same proteins in neurons as it does in yeast, amoebae, flies, and mammalian cell lines. Furthermore, there were 32 proteins enriched for cytoskeletal regulatory functions (Figure 1L), including actin-modulatory molecules such as the Arp2/3 complex subunit ARPC5, which is consistent with WASH's role in activating this complex to stimulate actin polymerization at endosomes for vesicular scission (Billadeau et al., 2010; Derivery et al., 2009). The WASH1-BioID2 isolated complex also contained 28 proteins known to localize to the excitatory post-synapse (Figure 1M). This included many scaffolding proteins, such as SHANK2-3 and DLGAP2-4 (Chen et al., 2011; Mao et al., 2015; Monteiro and Feng, 2017; Wan et al., 2011), as well as modulators of synaptic receptors such as SYNGAP1 and SHISA6 (Barnett et al., 2006; Clement et al., 2012; Kim et al., 2003; Klaassen et al., 2016), which was consistent with the idea that vesicular trafficking plays an important part

in synaptic function and regulation. Taken together, these results support a major endosomal trafficking role of the WASH complex in mouse brain.

SWIP^{P1019R} does not incorporate into the WASH complex, reducing its stability and levels *in vivo*. To determine how disruption of the WASH complex may lead to disease, we generated a mouse model of a human missense mutation found in children with intellectual disability, *WASHC4*^{c.3056c>g} (encoding protein SWIP^{P1019R}) (Ropers et al., 2011). Due to the sequence homology of human and mouse *Washc4* genes, we were able to introduce the same point mutation in exon 29 of murine *Washc4* using CRISPR (Derivery and Gautreau, 2010; Ropers et al., 2011). This C>G point mutation results in a Proline>Arginine substitution at position 1019 of SWIP's amino acid sequence (Figure 2A), a region thought to be critical for its binding to another WASH component, Strumpellin (Jia et al., 2010; Ropers et al., 2011). Western blot analysis of brain lysate from adult homozygous SWIP^{P1019R} mutant mice (referred to from here on as MUT mice) displayed significantly decreased expression of two WASH complex members, WASH1 and Strumpellin (Figure 2B). These results phenocopied data from the human patients (Ropers et al., 2011) and suggested that the WASH complex is unstable in the presence of this SWIP point mutation *in vivo*. To test whether this mutation disrupted interactions between WASH complex subunits, we compared the ability of wild-type SWIP (WT) and SWIP^{P1019R} (MUT) to co-immunoprecipitate with WASH1 and Strumpellin in HEK cells. Compared to WT, MUT SWIP co-immunoprecipitated significantly less WASH1 and Strumpellin (IP: 41.4% and 54.8% of WT SWIP, respectively), suggesting that the SWIP^{P1019R} mutation hinders WASH complex formation (Figure S1). Together these data

support the notion that SWIP^{P1019R} is a damaging mutation that not only impairs its function, but also results in significant reductions of the WASH complex as a whole.

Spatial proteomics and unbiased network covariation analysis reveal significant disruptions in the endo-lysosomal pathway of SWIP^{P1019R} mutant mouse brain.

Next, we aimed to understand the molecular impact of the SWIP^{P1019R} mutation on the subcellular organization of the brain proteome. We employed multiplex tandem-mass-tag (TMT) proteomics to quantify proteins in subcellular brain fractions. In addition to evaluating differential protein abundance between WT and SWIP^{P1019R} MUT brain, we analyzed network-level changes in groups of covarying proteins to better understand WASH's function and explore the cellular mechanisms by which SWIP^{P1019R} causes disease.

We performed a 16-plex TMT spatial proteomics experiment by isolating seven different cellular fractions from adult SWIP^{P1019R} MUT and WT mouse brain (Geladaki et al., 2019; Hallett et al., 2008). Brains from 10-month-old mice were gently homogenized to release intact organelles, followed by successive centrifugation steps to enrich subcellular compartments into different fractions based on their density (Figure 2C) (Geladaki et al., 2019). Seven WT and seven MUT fractions (each prepared from one brain, 14 samples total) were labeled with unique isobaric tandem-mass tags and concatenated. We also included two sample pooled quality controls (SPQCs), which allowed us to assess experimental variability and perform normalization between experiments. By performing this experiment in triplicate, deep coverage of the mouse brain proteome was obtained—across all 48 samples we quantified 86,551 peptides,

corresponding to 7,488 proteins. After data pre-processing, normalization, and filtering we retained 5,897 reproducibly quantified proteins in the final dataset (Table S2).

We used generalized linear models (GLMs) to assess differential protein abundance for intra-fraction comparisons between WT and MUT genotypes, and for overall comparisons between WT and MUT groups, adjusted for baseline differences in subcellular fraction. In the first analysis, there were 85 proteins with significantly altered abundance in at least one of the 7 subcellular fractions (Benjamini-Hochberg P-Adjust < 0.1, Table S2 and Figure S2A). Five proteins were differentially abundant between WT and MUT in all 7 fractions, including four WASH proteins and RAB21A—a known WASH interactor that functions in early endosomal trafficking (WASHC1, WASHC2, WASHC4, WASHC5, Figure 2E) (Olmo et al.; Simpson et al., 2004). The abundance of the remaining WASH complex protein, WASHC3, was found to be very low and was not retained in the final dataset due to its sparse quantification. These data affirm that the SWIP^{P1019R} mutation destabilizes the WASH complex. Next, to evaluate global differences between WT and MUT brain, we analyzed the average effect of genotype on protein abundance across all fractions. At this level, there were 687 differentially abundant proteins between WT and MUT brain (Bonferroni P-Adjust < 0.05) (Table S2). We then aimed to place these differentially abundant proteins into a more meaningful biological context using a systems-based approach.

For network-based analyses, we clustered the protein covariation network defined by pairwise correlations between all 5,897 proteins. Our data-driven, quality-based approach used Network Enhancement (Wang et al., 2018) to remove biological noise from the covariation network and employed the Leiden algorithm (Traag et al., 2019) to

identify optimal partitions of the graph. We enforced module quality by permutation testing (Ritchie et al., 2016) to ensure that identified modules exhibited a non-random topology. Clustering of the protein covariation graph identified 255 modules of proteins that strongly covaried together (see Methods for complete description of clustering approach).

To test for module-level differences between WT and MUT brain, we summarized modules for each biological replicate (a single subcellular fraction prepared from either a WT or MUT mouse) as the sum of their proteins, and extended our GLM framework to identify changes in module abundance (adjusted for fraction differences) between genotypes. 37 of the 255 modules exhibited significant differences in WT versus MUT brain (Bonferroni P-Adjust < 0.05; Table S3). Of note, the module containing the WASH complex, M19, was predicted to have endosomal function by annotation of protein function, and was enriched for proteins identified by WASH1-BioID2 (hypergeometric test P-Adjust < 0.05, bold node edges, Figure 2D). Similar to the WASH iBioID proteome (Figure 1), M19 contained components of the CCC (CCDC22, CCDC93, COMMD1-3, COMMD7, and COMMD9) and Retriever sorting complexes (VPS26C and VPS35L), but not the Retromer sorting complex, suggesting that in the brain, the WASH complex may not interact as closely with Retromer as it does in other cells (Figure 2D). Across all fractions, the abundance of M19 was significantly lower in MUT brain compared to WT, providing evidence that the SWIP^{P1019R} mutation reduces the stability of this protein subnetwork and impairs its function (Figure 2F-G).

In contrast to the observed decrease in abundance of M19, an endosomal module, we observed two modules, M213 and M2, which contained numerous key lysosomal components and exhibited increased abundance in MUT brain. M213 (Figure 3B)

248 contained the lysosomal protease Cathepsin B (CTSB) as well as two key lysosomal
249 hydrolases **GLB1** and **MAN2B**. M2 (Figure 3D) contained several lysosomal hydrolases
250 (e.g. GNS, **GLA**, and MAN2B1) as well as two Cathepsins (CTSS and CTSL). Notably,
251 M2 contains the lysosomal glycoprotein progranulin (GRN), a protein whose normal
252 function is integral to proper lysosome **function** and whose loss is widely linked with
253 neurodegenerative **path**ologies. M2 also contained the hydrolase and lysosomal storage
254 disorder gene, IDS—whose loss causes a metabolic disorder, Hunter’s Syndrome, whose
255 severe presentation includes a host of neurological symptoms. The overall increase in
256 abundance of modules M2 and M213 (Figure 3C and 3E) may reflect an increase in flux
257 through degradative lysosomal pathways in SWIP^{P1019R} brain (Figures 3B-3E). In support
258 of this hypothesis, other critical lysosomal proteins such as HEXB, were also increased
259 in MUT samples (Figure S2), indicating that disruption of the WASH complex increases
260 demand for lysosomal substrate degradation (Elia et al., 2019; Kuil et al., 2019; Mok et
261 al., 2003; Tiribuzi et al., 2011). Module 2 (Figure 3D) also included membrane proteins
262 ITGA5, a known WASH interactor, and the ATPase ATP13A2, a cation transporter whose
263 loss causes **Kufor-Rakeb** syndrome, a Parkinson’s disease, as well as excreted proteins
264 such as the metallopeptidase MMP17 that function in degradation and reorganization of
265 the extracellular matrix in normal and pathological conditions. The increased abundance
266 of these proteins in M2 may reflect their lysosomal degradation (Zech et al., 2011). Taken
267 together these changes seem to reflect a pathological condition characterized by changes
268 in lysosome metabolism and altered interactions with endosomal and cellular trafficking
269 machinery.

In addition to endo-lysosomal changes, network alterations were evident for an endoplasmic reticulum (ER) module (M83), supporting a shift in the proteostasis of mutant neurons (Figure S3C). Notably, within the ER module, M83, there was increased abundance of chaperones (e.g. HSPA5, PDIA3, PDIA4, PDIA6, and DNAJC3) that are commonly engaged in presence of misfolded proteins (Bartels et al., 2019; Kim et al., 2020; Montibeller and de Belleruche, 2018; Synofzik et al., 2014; Wang et al., 2016). This elevation of ER stress modulators can be indicative of neurodegenerative states, in which the unfolded protein response (UPR) is activated to resolve misfolded species (Garcia-Huerta et al., 2016; Hetz and Saxena, 2017). Taken together, these data demonstrate that the WASH complex functions in close proximity to endosomal compartments and that its loss from these locations likely alters not only endosomal trafficking, but also lysosomal function, with increased stress on cellular homeostasis.

Besides these endo-lysosomal and homeostatic changes, we also observed two synaptic modules (M35 and M143) that were also reduced in MUT brain (Figure S3D-E). These included mostly excitatory post-synaptic proteins such as SHISA6 (also identified in WASH1-BioID2, Figure 1), HOMER2, and DLG4, consistent with endosomal WASH influencing synaptic regulation. Decreased abundance of these modules indicates that loss of the WASH complex may result in failure of these proteins to be properly trafficked to the synapse.

Mutant neurons display structural abnormalities in endo-lysosomal compartments *in vitro*. Combined, the proteomics data strongly suggested that endo-lysosomal pathways are altered in adult SWIP^{P1019R} mutant mouse brain. Next, we analyzed whether

structural changes in this system were evident in primary neurons. Cortical neurons from littermate WT and MUT P0 pups were cultured for 15 days *in vitro* (DIV15, Figure 4A), then fixed and stained for established markers of early endosomes (Early Endosome Antigen 1, EEA1; Figures 4B and 4C) and lysosomes (Cathepsin D, CathD; Figures 4D and 4E). Reconstructed three-dimensional volumes of EEA1 and Cathepsin D puncta revealed that MUT neurons display larger EEA1+ somatic puncta than WT neurons (Figures 4G and 4J), but no difference in the total number of EEA1+ puncta (Figure 4F). This finding is consistent with a loss-of-function mutation, as loss of WASH activity prevents cargo scission from endosomes and leads to cargo accumulation (Bartuzi et al., 2016; Gomez et al., 2012). On the other hand, MUT neurons exhibited significantly less Cathepsin D+ puncta than WT neurons (Figure 4H), but the remaining puncta were significantly larger than those of WT neurons (Figures 4I and 4K). These data support the finding that the SWIP^{P1019R} mutation results in both molecular and morphological abnormalities in the endo-lysosomal pathway.

SWIP^{P1019R} mutant brains exhibit markers of abnormal endo-lysosomal structures and cell death *in vivo*. As there is strong evidence that dysfunctional endo-lysosomal trafficking and elevated ER stress are associated with neurodegenerative disorders, adolescent (P42) and adult (10 month-old, 10mo) WT and MUT brain tissue were analyzed for the presence of cleaved caspase-3, a marker of apoptotic pathway activation, in three brain regions (Boatright and Salvesen, 2003; Porter and Jänicke, 1999). Very little cleaved caspase-3 staining was present in WT and MUT mice at adolescence (Figures 5A, 5B, and S4). However, at 10mo, the MUT motor cortices

displayed significantly greater cleaved caspase-3 staining compared to age-matched WT littermate controls (Figures 5D-5F), but did not exhibit any significant difference in the number of total cells (Figures 5G-5K). Furthermore, this difference appeared to be selective for the motor cortex, as we did not observe significant differences in cleaved caspase-3 staining at either age for striatal or cerebellar regions (Figures S4). These data suggested that neurons of the motor cortex were particularly susceptible to disruption of the endo-lysosomal pathway, perhaps because long-range corticospinal projections require high fidelity of trafficking pathways (Blackstone et al., 2011; Slosarek et al., 2018; Wang et al., 2014).

To further examine the morphology of primary motor cortex neurons at a subcellular resolution, samples from age-matched 7-month-old WT and MUT mice (7mo, 3 animals each) were imaged by transmission electron microscopy (TEM). Strikingly, we observed large electron-dense inclusions in the cell bodies of MUT neurons (arrows, Figure 5M; pseudo-colored region, 5O). These dense structures were associated electron-lucent lipid-like inclusions (asterisk, Figure 5O), and were consistent with lipofuscin accumulation at lysosomal residual bodies (Poët et al., 2006; Valdez et al., 2017; Yoshikawa et al., 2002). Lipofuscin is a by-product of lysosomal breakdown of lipids, proteins, and carbohydrates, which naturally accumulates over time in non-dividing cells such as neurons (Höhn and Grune, 2013; Moreno-García et al., 2018; Terman and Brunk, 1998). However, excessive lipofuscin accumulation is thought to be detrimental to cellular homeostasis by inhibiting lysosomal function and promoting oxidative stress, often leading to cell death (Brunk and Terman, 2002; Powell et al., 2005). As a result, elevated lipofuscin is considered a biomarker of neurodegenerative disorders, including

Alzheimer's disease, Parkinson's disease, and Neuronal Ceroid Lipofuscinoses (Moreno-García et al., 2018). Therefore, the marked increase in lipofuscin area and number seen in MUT electron micrographs (Figures 5P and 5Q, respectively) is consistent with an increase in lysosomal breakdown of cellular material. Together these data indicate that SWIP^{P1019R} results in pathological lysosomal function that could lead to neurodegeneration.

SWIP^{P1019R} mutant mice display persistent deficits in cued fear memory recall. To observe the functional consequences of the SWIP^{P1019R} mutation, we next studied WT and MUT mouse behavior. Given that children with homozygous SWIP^{P1019R} point mutations display intellectual disability (Ropers et al., 2011) and SWIP^{P1019R} mutant mice exhibit endo-lysosomal disruptions implicated in neurodegenerative processes, behavior was assessed at two ages: adolescence (P40-50), and mid-late adulthood (5.5-6.5 mo). Interestingly, MUT mice performed equivalently to WT mice in episodic and working memory paradigms, including novel object recognition and Y-maze alternations (Figure S5). However, in a fear conditioning task, MUT mice displayed a significant deficit in cued fear memory (Figure 6). This task tests the ability of a mouse to associate an aversive event (a mild electric footshock) with a paired tone (Figure 6A). Freezing behavior of mice during tone presentation is attributed to hippocampal or amygdala-based fear memory processes (Goosens and Maren, 2001; Maren and Holt, 2000; Vazdarjanova and McGaugh, 1998). Forty-eight hours after exposure to the paired tone and footshock, MUT mice showed a significant decrease in conditioned freezing to tone presentation compared to their WT littermates (Figures 6B and 6C). To ensure that this difference was

not due to altered sensory capacities of MUT mice, we measured the startle response of mice to both electric foot shock and presented tones. In line with intact sensation, MUT mice responded comparably to WT mice in these tests (Figure S6). These data demonstrate that although MUT mice perceive footshock sensations and auditory cues, it is their memory of these paired events that is significantly impaired. Additionally, this deficit in fear response was evident at both adolescence and adulthood (top panels, and bottom panels, respectively, Figures 6B and 6C). These changes are consistent with the hypothesis that SWIP^{P109R} is the cause of cognitive impairments in humans.

SWIP^{P1019R} mutant mice exhibit surprising motor deficits that are confirmed in human patients. Because SWIP^{P1019R} results in endo-lysosomal pathology consistent with neurodegenerative disorders in the motor cortex, we next analyzed motor function of the mice over time. First, we tested the ability of WT and MUT mice to remain on a rotating rod for five minutes (Rotarod, Figures 7A-7C). At both adolescence and adulthood, the MUT mice performed markedly worse than WT littermate controls (Fig 7C). Mouse performance was not significantly different across trials, which suggested that this difference in retention time was not due to progressive fatigue, but more likely due to an overall difference in motor control (Mann and Chesselet, 2015).

To study the animals' movement at a finer scale, the gait of WT and MUT mice was also analyzed using a TreadScan system containing a high-speed camera coupled to a transparent treadmill (Figure 7D) (Beare et al., 2009). Interestingly, while the gait parameters of mice were largely indistinguishable across genotypes at adolescence, a striking difference was seen when the same mice were aged to adulthood (Figures 7E-

7G). In particular, MUT mice took slower (Figure 7E), longer strides (Figure 7F), stepping closer to the midline of their body (track width, Figure S7), and their gait symmetry was altered so that their strides were no longer perfectly out of phase (out of phase=0.5, Figure 7G). While these differences were most pronounced in the rear limbs (as depicted in Figure 7E-7G), the same trends were present in front limbs (Figure S7). These findings demonstrate that SWIP^{P1019R} results in a progressive motor function decline that was detectable by the rotarod task at adolescence, but which became more prominent with age, as both gait and strength functions deteriorated.

These marked motor findings prompted us to re-evaluate the original reports of human SWIP^{P1019R} patients (Ropers et al., 2011). While developmental delay or learning difficulties were the primary impetus for medical evaluation, all patients also exhibited motor symptoms (mean age = 10.4 years old, Figure 7H). The patients' movements were described as "clumsy" with notable fine motor difficulties, dysmetria, dysdiadochokinesia, and mild dysarthria on clinical exam (Figure 7H). Recent communication with the parents of these patients, who are now an average of 21 years old, revealed no notable symptom exacerbation. It is therefore possible that the SWIP^{P1019R} mouse model either exhibits differences from human patients or may predict future disease progression for these individuals, given that we observed significant worsening at 5-6 months old in mice (which is thought to be equivalent to ~30-35 years old in humans)(Dutta and Sengupta, 2016; Zhang et al., 2019).

DISCUSSION

Taken together, the data presented here support a mechanistic model whereby SWIP^{P1019R} causes a loss of WASH complex function, resulting in endo-lysosomal disruption and accumulation of neurodegenerative markers, such as upregulation of unfolded protein response modulators and lysosomal enzymes, as well as build-up of lipofuscin and cleaved caspase-3 over time. To our knowledge, this study provides the first mechanistic evidence of WASH complex impairment having direct and indirect organellar effects that lead to cognitive deficits and progressive motor impairments (Figure 8).

Using *in vivo* proximity-based proteomics in wild-type mouse brain, we identify that the WASH complex interacts with the CCC (COMMD9 and CCDC93) and Retriever (VPS35L) cargo selective complexes (Bartuzi et al., 2016; Singla et al., 2019). Interestingly, we did not find significant enrichment of the Retromer sorting complex, a well-known WASH interactor, suggesting that it may play a minor role in neuronal WASH-mediated cargo sorting (Figure 1). These data are supported by our TMT proteomics and covariation network analyses of SWIP^{P1019R} mutant brain, which clustered the WASH, CCC, and Retriever complexes together in M19, but not the Retromer complex (Figure 2). Systems-level protein covariation analyses also revealed that disruption of these WASH-CCC-Retriever interactions may have multiple downstream effects on the endosomal machinery, since endosomal modules displayed significant changes in SWIP^{P1019R} brain (including M19, Figure 2, as well as M14, Figure S3B), with corresponding decreases in the abundance of endosomal proteins including Retromer subunits (VPS29 and VPS35), associated sorting nexins (e.g. SNX17 and SNX27), known WASH interactors (e.g. RAB21 and FKBP15), and cargos (e.g. LRP1 and ITGA3)

(Figures S2 and S3) (Farfán et al., 2013; Fedoseienko et al., 2018; Halff et al., 2019; Harbour et al., 2012; McNally et al., 2017; Olmo et al.; Pan et al., 2010; Ye et al., 2020; Zimprich et al., 2011). While previous studies have indicated that Retromer and CCC influence endosomal localization of WASH (Harbour et al., 2012; Phillips-Krawczak et al., 2015; Singla et al., 2019), our findings of altered endosomal networks containing decreased Retromer, Retriever, and CCC protein levels in SWIP^{P1019R} mutant brain point to a possible feedback mechanism wherein WASH impacts the protein abundance and/or stability of these interactors. Future studies defining the hierarchical interplay between the WASH, Retromer, Retriever, and CCC complexes in neurons could provide clarity on how these mechanisms are organized.

In addition to highlighting the neuronal roles of WASH in CCC- and Retriever-mediated endosomal sorting, our proteomics approach also identified protein modules with increased abundance in SWIP^{P1019R} mutant brain. The proteins in these modules fell into two interesting categories: lysosomal enzymes and proteins involved in the endoplasmic reticulum (ER) stress response. Of note, some of the lysosomal enzymes with elevated levels in MUT brain (GRN, M2; GNS, M213; and HEXB, M6; Figures 3 and S2) are also implicated in lysosomal storage disorders, where they generally have decreased, rather than increased, function or expression (Mok et al., 2003; Ward et al., 2017; Zampieri et al., 2009). We speculate that loss of WASH function in our mutant mouse model may lead to increased accumulation of cargo and associated machinery at early endosomes (as seen in Figure 4, enlarged EEA1⁺ puncta), eventually overburdening the vesicles and triggering transition to late endosomes for subsequent fusion with degradative lysosomes (Figure 8). This would effectively increase delivery of endosomal

substrates to the lysosome compared to baseline, resulting in enlarged, overloaded lysosomal structures, and elevated demand for degradative enzymes. For example, since mutant neurons display increased lysosomal module protein abundance (Figure 3), and larger lysosomal structures (Figures 4 and 5), they may require higher quantities of progranulin (GRN, Figure 3; M2) for sufficient lysosomal acidification (Tanaka et al., 2017).

Our findings that SWIP^{P1019R} results in reduced WASH complex stability and function, which may ultimately drive lysosomal dysfunction, are supported by studies in non-mammalian cells. For example, expression of a dominant-negative form of WASH1 in amoebae impairs recycling of lysosomal V-ATPases (Carnell et al., 2011) and loss of WASH in *Drosophila* plasmocytes affects lysosomal acidification (Gomez et al., 2012; Nagel et al., 2017; Zech et al., 2011). Moreover, mouse embryonic fibroblasts lacking WASH1 display abnormal lysosomal morphologies, akin to the structures we observed in cultured SWIP^{P1019R} MUT neurons (Gomez et al., 2012).

In addition to lysosomal dysfunction, endoplasmic reticulum (ER) stress is commonly observed in neurodegenerative states, where accumulation of misfolded proteins disrupts cellular proteostasis (Cai et al., 2016; Hetz and Saxena, 2017; Montibeller and de Belleruche, 2018). This cellular strain triggers the adaptive unfolded protein response (UPR), which attempts to restore cellular homeostasis by increasing the cell's capacity to retain misfolded proteins within the ER, remedy misfolded substrates, and trigger degradation of persistently misfolded species. Involved in this process are ER chaperones that we identified as increased in SWIP^{P1019R} mutant brain including BiP (HSPA5), calreticulin (CALR), calnexin (CANX), and the protein disulfide isomerase

family members (PDIA1, PDIA4, PDIA6) (M83; Figure S3) (Garcia-Huerta et al., 2016). Many of these proteins were identified in the ER protein module found to be significantly altered in MUT mouse brain (M83), supporting a network-level change in the ER stress response (Figure S3C). One notable exception to this trend was endoplasmin (HSP90B1, M136), which exhibited significantly decreased abundance in SWIP^{P1019R} mutant brain (Table S2). This is surprising given that endoplasmin has been shown to coordinate with BiP in protein folding (Sun et al., 2019), however it may highlight a possible compensatory mechanism. Additionally, prolonged UPR can stimulate autophagic pathways in neurons, where misfolded substrates are delivered to the lysosome for degradation (Cai et al., 2016). These data highlight a relationship between ER and endo-lysosomal disturbances as an exciting avenue for future research.

It has become clear that preservation of the endo-lysosomal system is critical to neuronal function, as mutations in mediators of this process are implicated in neurological diseases such as Parkinson's disease, Huntington's disease, Alzheimer's disease, Frontotemporal Dementia, Neuronal Ceroid Lipofuscinoses (NCLs), and Hereditary Spastic Paraplegia (Baker et al., 2006; Connor-Robson et al., 2019; Edvardson et al., 2012; Follett et al., 2019; Harold et al., 2009; Mukherjee et al., 2019; Pal et al., 2006; Quadri et al., 2013; Seshadri et al., 2010; Tachibana et al., 2019; Valdmann et al., 2007). These genetic links to predominantly neurodegenerative conditions have supported the proposition that loss of endo-lysosomal integrity can have compounding effects over time and contribute to progressive disease pathologies. In particular, NCLs are lysosomal storage disorders primarily found in children, with heterogenous presentations and multigenic causations (Mukherjee et al., 2019). The majority of genes implicated in NCLs

affect lysosomal enzymatic function or transport of proteins to the lysosome (Mukherjee et al., 2019; Poët et al., 2006; Ramirez-Montealegre and Pearce, 2005; Yoshikawa et al., 2002). Most patients present with marked neurological impairments, such as learning disabilities, motor abnormalities, vision loss, and seizures, and have the unifying feature of lysosomal lipofuscin accumulation upon pathological examination (Mukherjee et al., 2019). While the human SWIP^{P1019R} mutation has not been classified as an NCL (Ropers et al., 2011), findings from our mutant mouse model suggest that loss of WASH complex function leads to phenotypes bearing strong resemblance to NCLs, including lipofuscin accumulation (Figures 4-7). As a result, our mouse model could provide the opportunity to study these pathologies at a mechanistic level, while also enabling preclinical development of treatments for their human counterparts.

Currently there is an urgent need for greater mechanistic investigations of neurodegenerative disorders, particularly in the domain of endo-lysosomal trafficking. Despite the continual increase in identification of human disease-associated genes, our molecular understanding of how their protein equivalents function and contribute to pathogenesis remains limited. Here we employ a systems-level analysis of proteomic datasets to uncover biological perturbations linked to SWIP^{P1019R}. We demonstrate the power of combining *in vivo* proteomics and systems network analyses with *in vitro* and *in vivo* functional studies to uncover relationships between genetic mutations and molecular disease pathologies. Applying this platform to study organellar dysfunction in other neurodegenerative disorders with protein trafficking defects may uncover convergent disease pathways.

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AUTHOR CONTRIBUTIONS

Conceptualization: JLC, TWAB, IHK, and SHS; Methodology: JLC, TWAB, IHK, and SHS; Investigation: JLC, TWAB, IHK, GW, TH, RV, ES, and AR; Resources: ES, RV, and SHS; Writing—Original Draft: JLC and SHS; Writing—Editing: JLC, TWAB, IHK, and SHS; Visualization: JLC; Funding Acquisition: JLC, TWAB, IHK, and SHS. All authors discussed the results and commented on the manuscript.

DECLARATION OF INTERESTS

The authors declare no competing interests.

FIGURE TITLES AND LEGENDS

Figure 1. Identification of the WASH complex proteome *in vivo* confirms a neuronal role in endosomal trafficking

(A) The WASH complex is composed of five subunits, *Washc1* (WASH1), *Washc2* (FAM21), *Washc3* (CCDC53), *Washc4* (SWIP), and *Washc5* (Strumpellin). Human mutations in these components are associated with spastic paraplegia (De Bot et al., 2013; Jahic et al., 2015; Valdmanis et al., 2007), Ritscher-Schinzel Syndrome (Elliott et al., 2013), and intellectual disability (Assoum et al., 2020; Ropers et al., 2011).

(B) A BioID2 probe was attached to the c-terminus of WASH1 and expressed under the human synapsin-1 (hSyn1) promoter in an AAV construct for *in vivo* BioID (iBioID). iBioID probes (WASH1-BioID2-HA, or negative control solubleBioID2-HA) were injected into wild-type mouse brain at P0 and allowed to express for two weeks. Subcutaneous biotin injections (24 mg/kg) were administered over seven days for biotinylation, and then brains were harvested for isolation and purification of biotinylated proteins. LC-MS/MS identified proteins significantly enriched in all three replicates of WASH1-BioID2 samples over soluble-BioID2 controls.

(C) Representative image of WASH1-BioID2-HA expression in a mouse coronal brain section (Cx = cortex, Hipp = hippocampus, Thal = thalamus). Scale bar, 1 mm.

(D) Representative image of WASH1-BioID2-HA expression in mouse cortex (inset from C). Individual panels show nuclei (DAPI, blue), AAV construct HA epitope (green), and

biotinylated proteins (Streptavidin, red). Merged image shows colocalization of HA and Streptavidin (yellow). Scale bar, 50 μ m.

(E) Representative image of WASH1-BioID2-HA expression in mouse hippocampus (inset from C). Scale bar, 50 μ m.

(F) Representative image of WASH1-BioID2-HA expression in mouse thalamus (inset from C). Scale bar, 50 μ m.

(G) iBioID identified known and unknown proteins interactors of the WASH complex in murine neurons. Nodes size represents protein abundance fold-enrichment over negative control (range: 3 to 181.7), solid grey edges delineate iBioID interactions between the WASHC1 probe (seen in yellow at the center) and identified proteins, dashed edges indicate known protein-protein interactions from HitPredict database (López et al., 2015).

(H-I) Clustergrams of:

(H) All five WASH complex proteins identified by iBioID.

(I) Previously reported WASH interactors (13/174), including the CCC and Retriever complexes.

(J) Endosomal trafficking proteins (23/174 proteins).

(K) Endocytic proteins (24/174).

(L) Proteins involved in cytoskeletal regulation (32/174), including Arp2/3 subunit ARPC5.

(M) Synaptic proteins (28/174). Clustergrams were annotated by hand and cross-referenced with Metascape (Zhou et al., 2019) GO enrichment of WASH1 proteome constituents over all proteins identified in the BioID experiment.

Figure 2. SWIP^{P1019R} does not incorporate into the WASH complex, reducing its stability and levels *in vivo*

(A) Mouse model of the human SWIP^{P1019R} missense mutation created using CRISPR. A C>G point mutation was introduced into exon29 of murine *Washc4*, leading to a P1019R amino acid substitution.

(B) Representative western blot and quantification of WASH components, Strumpellin and WASH1 (predicted sizes in kDa: 134 and 72, respectively), as well as loading control β -Tubulin (55kDa) from whole adult whole brain lysate prepared from SWIP WT (*Washc4*^{C/C}) and SWIP homozygous MUT (*Washc4*^{G/G}) mice. Bar plots show quantification of band intensities relative to WT (n=3 mice per genotype). Strumpellin (WT 100.0 \pm 5.2%, MUT 3.5 \pm 0.7%, $t_{2.1}$ =18.44, p=0.0024) and WASH1 (WT 100.0 \pm 3.8%, MUT 1.1 \pm 0.4%, $t_{2.1}$ =25.92, p=0.0013) were significantly decreased. Equivalent amounts of protein were analyzed in each condition (β -Tubulin: WT 100.0 \pm 8.2%, MUT 94.1 \pm 4.1%, U=4, p>0.99).

(C) Spatial TMT proteomics experimental design. 7 subcellular fractions were prepared from one WT and one MUT mouse (10mo). These samples, as well as two pooled quality control (QC) samples, were labeled with unique TMT tags and concatenated for simultaneous LC-MS/MS analysis. This experiment was repeated three times (3 WT and 3 MUT brains total).

(D) Protein module 19 (M19) contains subunits of the WASH, CCC, and Retriever complexes. Node size denotes its weighted degree centrality (~importance in module), purple color indicates proteins with altered abundance in MUT brain relative to WT, black node border denotes proteins identified in the WASH1-BioID proteome (Figure 1), black

edges indicate known protein-protein interactions, and grey-red edges denote the relative strength of protein covariation within a module (gray = weak, red = strong).

(E) Difference in normalized protein abundance for four WASH proteins found in M19 (SWIP: WT 6.28 ± 0.41 , MUT 4.89 ± 0.28 , $p=3.98 \times 10^{-28}$; WASH1: WT 6.41 ± 0.47 , MUT 4.65 ± 0.62 , $p=7.92 \times 10^{-18}$; FAM21: WT 6.29 ± 0.48 , MUT 5.29 ± 0.49 , $p=3.27 \times 10^{-16}$; Strumpellin: WT 7.85 ± 0.52 , MUT 6.59 ± 0.53 , $p=9.06 \times 10^{-25}$) and one control (Tubulin 4a: WT 8.57 ± 0.52 , MUT 8.52 ± 0.58 , $p>0.99$) across all three experimental replicates (n=3 independent experiments), presented as \log_2 (adjusted protein intensities).

(F) Normalized average intensities for every protein within M19 across all seven subcellular fractions analyzed. Teal lines delineate protein levels in WT samples, purple lines delineate protein levels in MUT samples (averaged across three experimental replicates). Bolded lines demarcate the fitted intensity values for WT and MUT proteins (n=3 independent experiments).

(G) Difference in M19 abundance, adjusted for fraction differences and presented as \log_2 (adjusted module abundance) (WT 13.21 ± 0.003 , MUT 12.99 ± 0.003 , $p=0.0007$; n=3 independent experiments).

Figure 3. Disruption of lysosomal protein networks in SWIP^{P1019R} mutant brain

(A) Simplified schematic of the endo-lysosomal pathway in neurons. (B) Network graph of module 213 (M213). M213 contains multiple lysosomal proteins that are increased in MUT brain. Network attributes: Node size denotes its weighted degree centrality (~importance in module), blue color indicates proteins with altered abundance in MUT brain relative to WT, black edges indicate known protein-protein interactions, and grey-

red edges denote the relative strength of protein covariation within a module (gray = weak, red = strong).

(C) The overall effect of genotype on M213 abundance (WT 12.17 ± 0.002 , MUT 12.33 ± 0.002 , $p=0.0037$).

(D) Module 2 (M2) contains multiple lysosomal proteins with increased abundance in MUT brain compared to WT, including CTSS, CTSL, GRN, IDS, MAN2B1. Network attributes depicted: same as in A.

(E) Difference in M2 module abundance, adjusted for fraction differences for all seven of the brain fractions analyzed, and presented as \log_2 (adjusted module intensity) (WT 13.74 ± 0.001 , MUT 13.85 ± 0.0009 , $p=0.0006$). ** $p<0.01$, *** $p<0.001$, empirical Bayes quasi-likelihood F-test with Bonferroni correction (C, E).

Figure 4. SWIP^{P1019R} mutant neurons display structural abnormalities in endo-lysosomal compartments *in vitro*

(A) Experimental design. Cortices were dissected from P0 pups, and neurons were dissociated and cultured on glass coverslips for 15 days. Cultures were fixed, stained, and imaged using confocal microscopy. 3D puncta volumes were reconstructed from z-stack images using Imaris software.

(B) Representative 3D reconstruction of WT DIV15 neuron stained for EEA1 (yellow) and MAP2 (magenta).

(C) Representative 3D reconstruction of MUT DIV15 neuron stained for EEA1 (yellow) and MAP2 (magenta).

657 (D) Representative 3D reconstruction of DIV15 WT neuron stained for Cathepsin D (cyan)
658 and MAP2 (magenta).

659 (E) Representative 3D reconstruction of DIV15 MUT neuron stained for Cathepsin D
660 (cyan) and MAP2 (magenta).

661 (F) Graph of the average number of EEA1+ volumes per soma in each image (WT 95.0
662 ± 5.5 , n=24 neurons; MUT 103.7 ± 3.7 , n=24 neurons; $t_{40.2}=1.314$, p=0.1961).

663 (G) Graph of the average EEA1+ volume size per soma shows larger EEA1+ volumes in
664 MUT neurons (WT 0.15 $\pm 0.01 \mu\text{m}^3$, n=24 neurons; MUT 0.30 $\pm 0.02 \mu\text{m}^3$, n=24 neurons;
665 U=50, p<0.0001).

666 (H) Graph of the average number of Cathepsin D+ volumes per soma illustrates less
667 Cathepsin D+ volumes in MUT neurons (WT 30.4 ± 1.4 , n=42; MUT 17.2 ± 0.9 , n=42;
668 $t_{71}=7.943$, p<0.0001).

669 (I) Graph of the average Cathepsin D+ volume size per soma demonstrates larger
670 Cathepsin D+ volumes in MUT neurons (WT 0.54 $\pm 0.02 \mu\text{m}^3$, n=42; MUT 0.69 ± 0.04
671 μm^3 , n=42; $t_{63}=3.701$, p=0.0005).

672 (J) Histogram of EEA1+ volumes illustrate differences in size distributions between MUT
673 and WT neurons.

674 (K) Histogram of CathD+ volumes show differences in size distributions between MUT
675 and WT neurons. Analyses included at least three separate culture preparations. Scale
676 bars, 5 μm (B-E). Data reported as mean \pm SEM, error bars are SEM. ***p<0.001,
677 ****p<0.0001, two-tailed t-tests or Mann-Whitney U test (G).

678

Figure 5. SWIP^{P1019R} mutant brains exhibit markers of abnormal endo-lysosomal structures and cell death *in vivo*

(A-B) Representative images of adolescent (P42) WT and MUT motor cortex stained with cleaved caspase-3 (CC3, green).

(C) Anatomical representation of mouse brain with motor cortex highlighted in blue, adapted from Allen Brain Atlas (Oh et al., 2014).

(D-E) Representative images of adult (10 mo) WT and MUT motor cortex stained with CC3 (green).

(H) Graph depicting the normalized percentage of DAPI+ nuclei that are positive for CC3 per image. No difference is seen at P42, but the amount of CC3+ nuclei is significantly higher in aged MUT mice (P42 WT $6.97 \pm 0.80\%$, P42 MUT $5.26 \pm 0.90\%$, 10mo WT $25.38 \pm 2.05\%$, 10mo MUT $44.01 \pm 1.90\%$, $H=74.12$, $p<0.0001$). We observed no difference in number of nuclei per image between genotypes.

(F, G, I, and J) DAPI co-stained images for (A, B, D, and E, respectively). Scale bar for (A-J), 15 μm .

(K) Representative transmission electron microscopy (TEM) image taken of soma from adult (7mo) WT motor cortex. Arrowheads delineate electron-dense lipofuscin material, Nuc = nucleus.

(L) Representative transmission electron microscopy (TEM) image taken of soma from adult (7mo) MUT motor cortex.

(M) Inset from (K) highlights lysosomal structure in WT soma. Pseudo-colored region depicts lipofuscin area, demarcated as L.

(N) Inset from (L) highlights large lipofuscin deposit in MUT soma (L, pseudo-colored region) with electron-dense and electron-lucent lipid-like (asterisk) components.

(O) Graph of areas of electron-dense regions of interest (ROI) shows increased ROI size in MUT neurons (WT $2.4 \times 10^5 \pm 2.8 \times 10^4 \text{ nm}^2$, n=50 ROIs; MUT $8.2 \times 10^5 \pm 9.7 \times 10^4 \text{ nm}^2$, n=75 ROIs; U=636, p<0.0001).

(P) Graph of the average number of presumptive lysosomes with associated electron-dense material reveals increased number in MUT samples (WT 3.14 ± 0.72 ROIs, n=14 images; MUT 10.86 ± 1.42 ROIs, n=14 images; U=17, p<0.0001). For (N) and (O), images were taken from multiple TEM grids, prepared from n=3 animals per genotype. Scale bar for all TEM images, 1 μm . Data reported as mean \pm SEM, error bars are SEM. ***p<0.001, ****p<0.0001, Kruskal-Wallis test (F), Mann-Whitney U test (O-P).

Figure 6. SWIP^{P1019R} mutant mice display persistent deficits in cued fear memory recall

(A) Experimental fear conditioning paradigm. After acclimation to a conditioning chamber, mice received a mild aversive 0.4mA footshock paired with a 2900Hz tone. 48 hours later, the mice were placed in a chamber with different tactile and visual cues. The mice acclimated for two minutes and then the 2900Hz tone was played (no footshock) and freezing behavior was assessed.

(B) Line graphs of WT and MUT freezing response during cued tone memory recall. Data represented as average freezing per genotype in 30 s time bins. The tone is presented after t = 120 s, and remains on for 120 seconds (Tone ON). Two different cohorts of mice were used for age groups P42 (top) and 6.5mo (bottom). Two-way ANOVA analysis of

average freezing during Pre-Tone and Tone periods reveal a Genotype x Time effect at P42 (WT n=10, MUT n=10, $F_{1,18}=4.944$, $p=0.0392$) and 6.5mo (WT n=13, MUT n=11, $F_{1,22}=13.61$, $p=0.0013$).

(C) Graphs showing the average %time freezing per animal before and during tone presentation. Top: freezing is reduced by 20% in MUT adolescent mice compared to WT littermates (Pre-tone WT $16.5 \pm 2.2\%$, n=10; Pre-tone MUT $13.0 \pm 1.8\%$, n=10; $t_{36}=0.8569$, $p=0.6366$; Tone WT $52.8 \pm 3.8\%$, n=10; Tone MUT $38.0 \pm 3.6\%$, n=10; $t_{36}=3.539$, $p=0.0023$), Bottom: freezing is reduced by over 30% in MUT adult mice compared to WT littermates (Pre-tone WT $21.1 \pm 2.7\%$, n=13; Pre-tone MUT $23.7 \pm 3.8\%$, n=11; $t_{44}=0.4675$, $p=0.8721$; Tone WT $69.7 \pm 4.3\%$, n=13; Tone MUT $53.1 \pm 5.2\%$, n=11; $t_{44}=2.921$, $p=0.0109$). Data reported as mean \pm SEM, error bars are SEM. * $p<0.05$, ** $p<0.01$, two-way ANOVAs (B) and Sidak's post-hoc analyses (C).

Figure 7. SWIP^{P1019R} mutant mice exhibit surprising motor deficits that are confirmed in human patients

(A) Rotarod experimental setup. Mice walked atop a rod rotating at 32rpm for 5 minutes, and the duration of time they remained on the rod before falling was recorded.

(B) Line graph of average duration animals remained on the rod per genotype across four trials, with an inter-trial interval of 40 minutes. The same cohort of animals was tested at two different ages, P45 (top) and 5.5 months (bottom). Genotype had a significant effect on task performance at both ages (top, P45: genotype effect, $F_{1,25}=7.821$, $p=0.0098$. bottom, 5.5mo: genotype effect, $F_{1,23}=7.573$, $p=0.0114$).

746 (C) Graphs showing the average duration each animal remained on the rod across trials.
747 At both ages, the MUT mice exhibited an almost 50% reduction in their ability to remain
748 on the rod (Top, P45: WT 169.9 ± 25.7 s, MUT 83.8 ± 15.9 s, $U=35$, $p=0.0054$. Bottom,
749 5.5mo: WT 135.9 ± 20.9 s, MUT 66.7 ± 9.5 s, $t_{18}=3.011$, $p=0.0075$).

750 (D) TreadScan task. Mice walked on a treadmill for 20 s while their gait was captured
751 with a high-speed camera. Diagrams of gait parameters measured in (E-G) are shown
752 below the TreadScan apparatus.

753 (E) Average swing time per stride for hindlimbs. At P45 (top), there is no significant
754 difference in rear swing time (WT 156.2 ± 22.4 ms, MUT 132.3 ± 19.6 ms, $U=83$,
755 $p=0.7203$). At 5.5mo (bottom), MUT mice display significantly longer rear swing time (WT
756 140 ± 6.2 ms, MUT 252.0 ± 21.6 ms, $t_{12}=4.988$, $p=0.0003$).

757 (F) Average stride length for hindlimbs. At P45 (top), there is no significant difference in
758 stride length (WT 62.3 ± 2.0 mm, MUT 60.5 ± 2.1 mm, $U=75$, $p=0.4583$). At 5.5mo
759 (bottom), MUT mice take significantly longer strides with their hindlimbs (WT 60.8 ± 0.8
760 mm, MUT 73.6 ± 2.7 mm, $t_{11.7}=4.547$, $p=0.0007$).

761 (G) Average homologous coupling for front and rear limbs. Homologous coupling is 0.5
762 when the left and right feet are completely out of phase. At P45 (top), WT and MUT mice
763 exhibit normal homologous coupling (WT 0.48 ± 0.005 , MUT 0.48 ± 0.004 , $U=76.5$,
764 $p=0.4920$). At 5.5 mo (bottom), MUT mice display decreased homologous coupling,
765 suggestive of abnormal gait symmetry (WT 0.48 ± 0.003 , MUT 0.46 ± 0.004 , $t_{18.8}=3.715$,
766 $p=0.0015$). At P45: $n=14$ WT, $n=13$ MUT; At 5.5mo: $n=14$ WT, $n=11$ MUT.

767 (H) Table of motor findings in clinical exam of human patients with homozygous
768 SWIP^{P1019R} mutation. All patients exhibit motor dysfunction (+ = symptom present). Data

reported as mean \pm SEM, error bars are SEM. * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$, **** $p < 0.0001$, two-way repeated measure ANOVAs (B), Mann-Whitney U tests and two-tailed t-tests (C-G).

Figure 8. Model of neuronal endo-lysosomal pathology in SWIP^{P1019R} mutant mice

(A) Wild-type WASH function in mouse brain. Under normal conditions, the WASH complex interacts with many endosomal proteins and cytoskeletal regulators, such as the Arp2/3 complex. These interactions enable restructuring of the endosome surface (actin in gray) and allow for cargo segregation and scission of vesicles. Substrates are transported to the late endosome for lysosomal degradation, to the Golgi network for modification, or to the cell surface for recycling.

(B) Loss of WASH function leads to increased lysosomal degradation in mouse brain. Destabilization of the WASH complex leads to enlarged endosomes and lysosomes, with increased substrate accumulation at the lysosome. This suggests an increase in flux through the endo-lysosomal pathway, possibly as a result of mis-localized endosomal substrates.

(C) Wild-type mice exhibit normal motor function.

(D) SWIP^{P1019R} mutant mice display progressive motor dysfunction in association with these subcellular alterations.

STAR★METHODS

Detailed methods are provided in the online version of this paper and include the following:

- **KEY RESOURCES TABLE**

- 792 • **RESOURCE AVAILABILITY**
- 793 ○ **Lead Contact**
- 794 ○ **Materials Availability**
- 795 ○ **Data and Code Availability**
- 796 • **EXPERIMENTAL MODEL AND SUBJECT DETAILS**
- 797 ○ **Animals**
- 798 ○ **Cell Culture**
- 799 ○ **Primary Neuronal Culture**
- 800 ○ **Human Subjects**
- 801 • **METHOD DETAILS**
- 802 ○ **Plasmid DNA Constructs**
- 803 ○ **AAV Viral Preparation**
- 804 ○ **Immunocytochemistry**
- 805 ○ **Immunohistochemistry**
- 806 ○ **Western Blotting**
- 807 ○ **Immunoprecipitation**
- 808 ○ **Electron Microscopy**
- 809 ○ **iBioID Sample Preparation**
- 810 ○ **LC-MS/MS for iBioID**
- 811 ○ **LOPIT-DC Subcellular Fractionation**
- 812 ○ **16-plex TMT LC-MS/MS**
- 813 ○ **Mouse Behavioral Assays**
- 814 ▪ **Y-maze**

- 815 ▪ **Novel Object Recognition**
- 816 ▪ **TreadScan**
- 817 ▪ **Steady Speed Rotarod**
- 818 ▪ **Fear Conditioning**
- 819 ▪ **Hearing Test**
- 820 ▪ **Startle Response (Somatosensation)**
- 821 • **QUANTIFICATION AND STATISTICAL ANALYSIS**
- 822 ○ **Cleaved Caspase-3 Image Analysis**
- 823 ○ **Imaris 3D Reconstruction**
- 824 ○ **iBioID Quantitative Analysis**
- 825 ○ **TMT Proteomics Quantitative Analysis**

827 **KEY RESOURCES TABLE**

828

829 **RESOURCE AVAILABILITY**

830 **Lead Contact**

831 Further information and requests for resources and reagents should be directed to and
832 will be fulfilled by the Lead Contact, Scott Soderling (scott.soderling@duke.edu).

833 **Materials Availability**

834 Plasmids generated by this study are available upon request from corresponding author
835 Scott H. Soderling (scott.soderling@duke.edu).

836 **Data and Code Availability**

The data and source code used in this study are available online at <https://github.com/twesleyb/SwipProteomics>.

EXPERIMENTAL MODEL AND SUBJECT DETAILS

Animals

We generated *Washc4* mutant (SWIP^{P1019R}) mice in collaboration with the Duke Transgenic Core Facility to mimic the *de novo* human variant at amino acid 1019 of human *WASHc4*. A CRISPR-induced CCT>CGT point mutation was introduced into exon 29 of *Washc4*. 50ng/μl of the sgRNA (5'-ttgagaataactcacaagaggagg-3'), 100ng/μl Cas9 mRNA, and 100ng/μl of a repair oligo containing the C>G mutation were injected into the cytoplasm of B6SJLF1/J mouse embryos (Jax #100012) (see Supplemental Table 4 for the sequence of the repair oligo). Mice with germline transmission were then backcrossed into a C57BL/6J background (Jax #000664). At least 5 backcrosses were obtained before animals were used for behavior. We bred heterozygous SWIP^{P1019R} mice together to obtain age-matched mutant and wild-type genotypes for cell culture and behavioral experiments. Genetic sequencing was used to screen for germline transmission of the C>G point mutation (*FOR*: 5'-tgctttagatgttttcct-3', *REV*: 5'-gttaacatgatcctatggcg-3'). All mice were housed in the Duke University's Division of Laboratory Animal Resources or Behavioral Core facilities at 2-5 animals/cage on a 14:10h light:dark cycle. All experiments were conducted with a protocol approved by the Duke University Institutional Animal Care and Use Committee in accordance with NIH guidelines.

Cell Culture

HEK293T cells (ATCC #CRL-11268) were purchased from the Duke Cell Culture facility. and were tested for mycoplasma contamination. HEK239T cells were used for coimmunoprecipitation experiments and preparation of AAV viruses.

Primary Neuronal Culture

Primary neuronal cultures were prepared from P0 mouse cortex. P0 mouse pups were rapidly decapitated and cortices were dissected and kept individually in 5ml Hibernate A (Thermo #A1247501) supplemented with 2% B27(Thermo #17504044) at 4°C overnight to allow for individual animal genotyping before plating. Neurons were then treated with Papain (Worthington #LS003120) and DNase (VWR #V0335)-supplemented Hibernate A for 18min at 37°C and washed twice with plating medium (plating medium: Neurobasal A (Thermo #10888022) supplemented with 10% horse serum, 2% B-27, and 1% GlutaMAX (Thermo #35050061)), and triturated before plating at 250,000 cells/well on poly-L-lysine-treated coverslips (Sigma #P2636) in 24-well plates. Plating medium was replaced with growth medium (Neurobasal A, 2% B-27, 1% GlutaMAX) 2 hours later. Cell media was supplemented and treated with AraC at DIV5 (5uM final concentration/well). Half-media changes were then performed every 4 days.

Human Subjects

We retrospectively analyzed clinical findings from seven children with homozygous *WASHC4*^{c.3056C>G} mutations (obtained by Dr. Rajab in 2010 at the Royal Hospital, Muscat, Oman). The original report of these human subjects and consent for data use can be found in (Ropers et al., 2011).

METHOD DETAILS

Plasmid DNA Constructs

For immunoprecipitation experiments, a pmCAG-SWIP-WT-HA construct was generated by PCR amplification of the human *WASHC4* sequence, which was then inserted between NheI and Sall restriction sites of a pmCAG-HA backbone generated in our lab. Site-directed mutagenesis (Agilent #200517) was used to introduce a C>G point mutation into this pmCAG-SWIP-WT-HA construct for generation of a pmCAG-SWIP-MUT-HA construct (*FOR*: 5'-ctacaaagttgagggtcagacggggaacaattatatagaaa-3', *REV*: 5'-tttctatataattgttccccgtctgaccctcaactttgttag-3'). For iBioID experiments, an AAV construct expressing hSyn1-WASH1-BioID2-HA was generated by cloning a *Washc1* insert between Sall and HindIII sites of a pAAV-hSyn1-Actin Chromobody-Linker-BioID2-pA construct (replacing Actin Chromobody) generated in our lab. This backbone included a 25nm GS linker-BioID2-HA fragment from Addgene #80899, generated by Roux et al (Kim et al., 2016). An hSyn1-solubleBioID2-HA construct was created similarly, by removing Actin Chromobody from the above construct. Oligonucleotide sequences are reported in Supplementary Table 4. Sequences of the plasmid DNA constructs are available online (see Key Resources Table).

AAV Viral Preparation

AAV preparations were performed as described previously (Uezu et al., 2016). The day before transfection, HEK293T cells were plated at a density of 1.5×10^7 cells per 15cm² plate in DMEM media with 10% fetal bovine serum and 1% Pen/Strep (Thermo #11965-092, Sigma #F4135, Thermo #15140-122). Six HEK293T 15cm² plates were used per viral preparation. The next day, 30µg of pAd-DeltaF6 helper plasmid, 15µg of AAV2/9 plasmid, and 15µg of an AAV plasmid carrying the transgene of interest were

905 mixed in OptiMEM with PEI-MAX (final concentration 80µg/ml, Polysciences #24765).
906 2ml of this solution were then added dropwise to each of the 6 HEK293T 15cm² plates.
907 Eight hours later, the media was replaced with 20ml DMEM+10%FBS. 72 hours post-
908 transfection, cells were scraped and collected in the media, pooled, and centrifuged at
909 1,500rpm for 5min at RT. The final pellet from the 6 cell plates was resuspended in 5ml
910 of cell lysis buffer (15 mM NaCl, 5 mM Tris-HCl, pH 8.5), and freeze-thawed three times
911 using an ethanol/dry ice bath. The lysate was then treated with 50U/ml of Benzonase
912 (Novagen #70664), for 30min in a 37°C water bath, vortexed, and then centrifuged at
913 4,500rpm for 30min at 4°C. The resulting supernatant containing AAV particles was added
914 to the top of an iodixanol gradient (15%, 25%, 40%, 60% top to bottom) in an Optiseal
915 tube (Beckman Coulter #361625). The gradient was then centrifuged using a Beckman
916 Ti-70 rotor in a Beckman XL-90 ultracentrifuge at 67,000rpm for 70min, 18°C. The purified
917 viral solution was extracted from the 40%/60% iodixanol interface using a syringe, and
918 placed into an Amicon 100kDa filter unit (#UFC910024). The viral solution was washed
919 in this filter 3 times with 1X ice-cold PBS by adding 5ml of PBS and centrifuging at
920 4,900rpm for 45min at 4°C to obtain a final volume of approximately 200µl of concentrated
921 virus that was aliquoted into 5-10µl aliquots and stored at -80°C until use.

922 **Immunocytochemistry**

923 Primary antibodies: Rabbit anti-EEA1 (Cell Signaling Technology #C45B10,
924 1:500), Rat anti-CathepsinD (Novus #204712, 1:250), Guinea Pig anti-MAP2 (Synaptic
925 Systems #188004, 1:500)

926 Secondary antibodies: Goat anti-Rabbit Alexa Fluor 568 (Invitrogen #A11036,
927 1:1000), Goat anti-Guinea Pig Alexa Fluor 488 (Invitrogen #A11073, 1:1000), Goat anti-

928 Rat Alexa Fluor 488 (Invitrogen #A11006, 1:1000), Goat anti-Guinea Pig Alexa Fluor 555
929 (Invitrogen #A21435, 1:1000)

930 At DIV15, neurons were fixed for 15 minutes using ice-cold 4%PFA/4% sucrose in
931 1X PBS, pH 7.4 (for EEA1 staining), or 30 minutes with 50% Bouin's solution/4% sucrose
932 (for CathepsinD staining, Sigma #HT10132), pH 7.4(Cheng et al., 2018). Fixed neurons
933 were washed with 1X PBS, then permeabilized with 0.25% TritonX-100 in PBS for 8
934 minutes at room temperature (RT), and blocked with 5%normal goat serum/0.2%Triton-
935 X100 in PBS (blocking buffer) for 1 hour at RT with gentle rocking. For EEA1/MAP2
936 staining, samples were incubated with primary antibodies diluted in blocking buffer at RT
937 for 1 hour. For CathepsinD/MAP2 staining, samples were incubated with primary
938 antibodies diluted in blocking buffer overnight at 4°C. For both conditions, samples were
939 washed three times with 1X PBS, and incubated for 30min at RT with secondary
940 antibodies, protected from light. After secondary antibody staining, coverslips were
941 washed three times with 1X PBS, and mounted with FluoroSave mounting solution
942 (Sigma #345789). See antibody section for primary and secondary antibody
943 concentrations.

944 **Immunohistochemistry**

945 Primary antibodies: Rabbit anti-Cleaved Caspase-3 (Cell Signaling Technology
946 #9661, 1:2000), Mouse anti-Calbindin (Sigma #C9848, 1:2000), Rat anti-HA 3F10 (Sigma
947 #12158167001, 1:500)

948 Secondary antibodies: Donkey anti-Rabbit Alexa Fluor 488 (Invitrogen #A21206,
949 1:2000), Goat anti-Mouse Alexa Fluor 594 (Invitrogen #A11032, 1:2000), Goat anti-Rat
950 Alexa Fluor 488 (Invitrogen #A11006, 1:5000), Streptavidin Alexa Fluor 594 conjugate

951 (Invitrogen #S32356, 1:5000), 4',6-diamidino-2-phenylindole (DAPI, Sigma #D9542,
952 1:1000 for 10min at RT)

953 Mice were deeply anesthetized with isoflurane and then transcardially perfused
954 with ice-cold heparinized PBS (25U/ml) by gravity flow. After clearing of liver and lungs
955 (~2min), perfusate was switched to ice-cold 4% PFA in 1X PBS (pH 7.4) for 15 minutes.
956 Brains were dissected, post-fixed in 4%PFA overnight at 4°C, and then cryoprotected in
957 30% sucrose/1X PBS for 48hr at 4°C. Fixed brains were then mounted in OTC (Sakura
958 TissueTek #4583) and stored at -20°C until cryosectioning. Every third sagittal section
959 (30 µm thickness) was collected from the motor cortex and striatal regions. Free-floating
960 sections were then permeabilized with 1%TritonX-100 in 1X PBS at RT for 2 hr, and
961 blocked in 1X blocking solution (Abcam #126587) diluted in 0.2%TritonX-100 in 1X PBS
962 for 1hr at RT. Sections were then incubated in primary antibodies diluted in the 1X
963 blocking solution for two overnights at 4°C. After three washes with 0.2%TritonX-100 in
964 1X PBS, the sections were then incubated in secondary antibodies diluted in 1X blocking
965 buffer for one overnight at 4°C. Sections were then washed four times with 0.2%TritonX-
966 100 in 1X PBS at RT, and mounted onto coverslips with FluoroSave mounting solution
967 (Sigma #345789).

968 **Western Blotting**

969 Primary antibodies: Rabbit anti-Strumpellin (Santa Cruz #sc-87442, 1:500), Rabbit
970 anti-WASH1 c-terminal (Sigma #SAB4200373, 1:500), Mouse anti-Beta Tubulin III
971 (Sigma #T8660, 1:10,000), Mouse anti-HA (BioLegend #MMS-101P, 1:5000)

972 Secondary antibodies: Donkey anti-Rabbit-HRP (GE Life Sciences #NA934,
973 1:5,000), Goat anti-mouse-HRP (GE Life Sciences #NA931, 1:5000)

Ten micrograms of each sample were electrophoresed through a 12-well, 4-20% SDS-PAGE gel (Bio-Rad #4561096) at 100V for 1hr at RT, transferred onto a nitrocellulose membrane (GE Life Sciences #GE10600002) at 100V for 70min at RT on ice, and blocked with 5% nonfat dry milk in TRIS-buffered saline containing 0.05% Tween-20 (TBST, pH 7.4). Gels were saved for Coomassie staining at RT for 30 min. Membranes were probed with one primary antibody at a time for 24hr at 4°C, then washed four times with TBST at RT before incubating with the corresponding species-specific secondary antibody at RT for 1hr. Membranes were washed with TBST, and then enhanced chemiluminescence (ECL) substrate was added (Thermo Fischer #32109). Membranes were exposed to autoradiography films and scanned with an Epson 1670 at 600dpi. We probed each membrane with one antibody at a time, then stripped the membrane with stripping buffer (Thermo Fischer #21059) for 10min at RT, and then blocked for 1hr at RT before probing with the next antibody. Order of probes: Strumpellin, then β -Tubulin, then WASH1. We determined the optical density of the bands using Image J software (NIH). Data obtained from three independent experiments were plotted and statistically analyzed using GraphPad Prism (version 8) software.

Immunoprecipitation

HEK293T cells were transfected with pmCAG-SWIP-WT-HA or pmCAG-SWIP-MUT-HA constructs for three days, as previously described (Mason et al., 2011). Cells were lysed with lysis buffer (25mM HEPES, 150mM NaCl, 1mM EDTA, 1% NonidetP-40, pH 7.4) containing protease inhibitors (5mM NaF, 1mM orthovanadate, 1mM AEBSF, and 2 μ g/mL leupeptin/pepstatin) and centrifuged at 1,700g for 5 min. Collected supernatant was incubated with 30 μ l of pre-washed anti-HA agarose beads (Sigma #A2095) on a

sample rotator (15 rpm) for 2 hrs at 4°C. Beads were then washed 3 times with lysis buffer, and sample buffer was added before subjecting to immunoblotting as described above. The protein-transferred membrane was probed individually for WASH1, Strumpellin, and HA. Data were collected from four separate preparations of WT and MUT conditions.

Electron Microscopy

Adult (7mo) WT and MUT SWIP^{P1019R} mice were deeply anesthetized with isoflurane and then transcardially perfused with warmed heparinized saline (25U/ml heparin) for 4 minutes, followed by ice-cold 0.15M cacodylate buffer pH 7.4 containing 2.5% glutaraldehyde (Electron Microscopy Sciences #16320), 3% paraformaldehyde, and 2mM CaCl₂ for 15 minutes. Brain samples were dissected and stored on ice in the same fixative for 2 hours before washing in 0.1M sodium cacodylate buffer (3 changes for 15 minutes each). Samples were then post-fixed in 1.0% OsO₄ in 0.1 M Sodium cacodylate buffer for 1 hour on a rotator. Samples were then washed in 3, 15-minute changes of 0.1M sodium cacodylate. Samples were then placed into *en bloc* stain (1% uranyl acetate) overnight at 4°C. Subsequently, samples were dehydrated in a series of ascending acetone concentrations including 50%, 70%, 95%, and 100% for three cycles with 15 minutes incubation at each concentration change. Samples were then placed in a 50:50 mixture of epoxy resin (Epon) and acetone overnight on a rotator. This solution was then replaced twice with 100% fresh Epon for at least 2 hours at room temperature on a rotator. Samples were embedded with 100% Epon resin in BEEM capsules (Ted Pella) for 48 hours at 60°C. Samples were ultrathin sectioned to 60-70nm on a Reichert Ultracut E ultramicrotome. Harvested grids were then stained with 2% uranyl acetate in

50% ethanol for 30 minutes and Sato's lead stain for 1 min. Micrographs were acquired using a Phillips CM12 electron microscope operating at 80Kv, at 1700x magnification. Micrographs were analyzed in Adobe Photoshop 2019, using the "magic wand" tool to demarcate and measure the area of electron-dense and electron-lucent regions of interest (ROIs). Statistical analyses of ROI measurements were performed in GraphPad Prism (version 8) software. The experimenter was blinded to genotype for image acquisition and analysis.

iBioID Sample Preparation

AAV2/9 viral probes, hSyn1-WASH1-BioID2-HA or hSyn1-solubleBioID2-HA, were injected into wild-type CD1 mouse brains using a Hamilton syringe (#7635-01) at age P0-P1 to ensure viral spread throughout the forebrain (Glascocock et al., 2011). 15 days post-viral injection, biotin was subcutaneously administered at 24mg/kg for seven consecutive days for biotinylation of proteins in proximity to BioID2 probes. Whole brains were extracted on the final day of biotin injections, snap frozen, and stored in liquid nitrogen until protein purification. Seven brains were used for protein purification of each probe, and each purification was performed three times independently (21 brains total for WASH1-BioID2, 21 for solubleBioID2).

We performed all homogenization and protein purification on ice. A 2ml Dounce homogenizer was used to individually homogenize each brain in a 1:1 solution of Lysis-R:2X-RIPA buffer solution with protease inhibitors (Roche cOmplete tablets #11836153001). Each sample was sonicated three times for 7 seconds and then centrifuged at 5000g for 5min at 4°C. Samples were transferred to Beckman Coulter 1.5ml tubes (#344059), and then spun at 45,000rpm in a Beckman Coulter tabletop

ultracentrifuge (TLA-55 rotor) for 1hr at 4°C. SDS was added to supernatants (final 1%) and samples were then boiled for 5min at 95°C. We next combined supernatants from the same condition together (WASH1-BioID2 vs. solubleBioID2) in 15ml conical tubes to rotate with 30µl high-capacity NeutrAvidin beads overnight at 4°C (Thermo #29204).

The following day, all steps were performed under a hood with keratin-free reagents. Samples were spun down at 6000rpm, 4°C for 5min to pellet the beads and remove supernatant. The pelleted beads then went through a series of washes, each for 10 min at RT with 500ul of solvent, and then spun down on a tabletop centrifuge to pellet the beads for the next wash. The washes were as follows: 2% SDS twice, 1% TritonX100-1%deoxycholate-25mM LiCl₂ once, 1M NaCl twice, 50mM Ammonium Bicarbonate (Ambic) five times. Beads were then mixed 1:1 with a 2X Laemmli sample buffer that contained 3mM biotin/50mM Ambic, boiled for 5 mins at 95°C, vortexed three times, and then biotinylated protein supernatants were stored at -80°C until LC-MS/MS.

LC-MS/MS for iBioID

We gave the Duke Proteomics and Metabolomics Shared Resource (DPMSR) six eluents from streptavidin resins (3 x WASH1-BioID2, 3 x solubleBioID2), stored on dry ice. Samples were reduced with 10 mM dithiothreitol for 30 min at 80°C and alkylated with 20 mM iodoacetamide for 30 min at room temperature. Next, samples were supplemented with a final concentration of 1.2% phosphoric acid and 256 µL of S-Trap (Protifi) binding buffer (90% MeOH/100mM TEAB). Proteins were trapped on the S-Trap, digested using 20 ng/µl sequencing grade trypsin (Promega) for 1 hr at 47°C, and eluted using 50 mM TEAB, followed by 0.2% FA, and lastly using 50% ACN/0.2% FA. All samples were then lyophilized to dryness and resuspended in 20 µL 1%TFA/2%

acetonitrile containing 25 fmol/ μ L yeast alcohol dehydrogenase (UniProtKB P00330; ADH_YEAST). From each sample, 3 μ L was removed to create a pooled QC sample (SPQC) which was run analyzed in technical triplicate throughout the acquisition period.

Quantitative LC/MS/MS was performed on 2 μ L of each sample, using a nanoAcquity UPLC system (Waters) coupled to a Thermo QExactive HF-X high resolution accurate mass tandem mass spectrometer (Thermo) via a nanoelectrospray ionization source. Briefly, the sample was first trapped on a Symmetry C18 20 mm \times 180 μ m trapping column (5 μ L/min at 99.9/0.1 v/v water/acetonitrile), after which the analytical separation was performed using a 1.8 μ m Acquity HSS T3 C18 75 μ m \times 250 mm column (Waters) with a 90-min linear gradient of 5 to 30% acetonitrile with 0.1% formic acid at a flow rate of 400 nanoliters/minute (nL/min) with a column temperature of 55°C. Data collection on the QExactive HF-X mass spectrometer was performed in a data-dependent acquisition (DDA) mode of acquisition with a $r=120,000$ (@ m/z 200) full MS scan from m/z 375 – 1600 with a target AGC value of $3e^6$ ions followed by 30 MS/MS scans at $r=15,000$ (@ m/z 200) at a target AGC value of $5e^4$ ions and 45 ms. A 20s dynamic exclusion was employed to increase depth of coverage. The total analysis cycle time for each sample injection was approximately 2 hours.

LOPIT-DC Subcellular Fractionation

We performed three independent fractionation experiments with one adult SWIP mutant brain and one WT mouse brain fractionated in each experiment. Each mouse was sacrificed by isoflurane inhalation and its brain was immediately extracted and placed into a 2ml Dounce homogenizer on ice with 1ml isotonic TEVP homogenization buffer (320mM sucrose, 10mM Tris base, 1mM EDTA, 1mM EGTA, 5mM NaF, pH7.4)(Hallett et al.,

2008)). A cOmplete mini protease inhibitor cocktail tablet (Sigma #11836170001) was added to a 50ml TEVP buffer aliquot immediately before use. Brains were homogenized for 15 passes with a Dounce homogenizer to break the tissue, and then this lysate was brought up to a 5ml volume with additional TEVP buffer. Lysates were then passed through a 0.5ml ball-bearing homogenizer for two passes (14 μ m ball, Isobiotec) to release organelles. Final brain lysate volumes were approximately 7.5ml each. Lysates were then divided into replicate microfuge tubes (Beckman Coulter #357448) to perform differential centrifugation, following Geladaki et. al's LOPIT-DC protocol(Geladaki et al., 2019). Centrifugation was carried out at 4°C in a tabletop Eppendorf 5424 centrifuge for spins at 200g, 1,000g, 3,000g, 5,000g, 9,000g, 12,000g, and 15,000g. To isolate the final three fractions, a tabletop Beckman TLA-100 ultracentrifuge with a TLA-55 rotor was used at 4°C with speeds of: 30,000g, 79,000g, and 120,000g, respectively. Samples were kept on ice at all times and pellets were stored at -80°C. Pellets from seven fractions (5,000g-120,000g) were used for proteomic analyses.

16-plex TMT LC-MS/MS

The Duke Proteomics and Metabolomics Shared Resource (DPMSR) processed and prepared fraction pellets from all 42 frozen samples simultaneously (7 fractions per brain from 3 WT and 3 MUT brains). Due to volume constraints, each sample was split into 3 tubes, for a total of 126 samples, which were processed in the following manner: 100 μ L of 8M Urea was added to the first aliquot then probe sonicated for 5 seconds with an energy setting of 30%. This volume was then transferred to the second and then third aliquot after sonication in the same manner. All tubes were centrifuged at 10,000g and any residual volume from tubes 1 and 2 were added to tube 3. Protein concentrations

1112 were determined by BCA on the supernatant in duplicate (5 μ L each assay). Total protein
1113 concentrations for each replicate ranged from 1.1 mg/mL to 7.8 mg/mL with total protein
1114 quantities ranging from 108.3 to 740.81 μ g. 60 μ g of each sample was removed and
1115 normalized to 52.6 μ L with 8M Urea and 14.6 μ L 20% SDS. Samples were reduced with
1116 10 mM dithiothreitol for 30 min at 80°C and alkylated with 20 mM iodoacetamide for 30
1117 min at room temperature. Next, they were supplemented with 7.4 μ L of 12% phosphoric
1118 acid, and 574 μ L of S-Trap (Protifi) binding buffer (90% MeOH/100mM TEAB). Proteins
1119 were trapped on the S-Trap, digested using 20 ng/ μ L sequencing grade trypsin (Promega)
1120 for 1 hr at 47°C, and eluted using 50 mM TEAB, followed by 0.2% FA, and lastly using
1121 50% ACN/0.2% FA. All samples were then lyophilized to dryness.

1122 Each sample was resuspended in 120 μ L 200 mM triethylammonium bicarbonate,
1123 pH 8.0 (TEAB). From each sample, 20 μ L was removed and combined to form a pooled
1124 quality control sample (SPQC). Fresh TMTPro reagent (0.5 mg for each 16-plex reagent)
1125 was resuspended in 20 μ L 100% acetonitrile (ACN) and was added to each sample.
1126 Samples were incubated for 1 hour at RT. After the 1-hour reaction, 5 μ L of 5%
1127 hydroxylamine was added and incubated for 15 minutes at room temperature to quench
1128 the reaction. Each 16-plex TMT experiment consisted of the WT and MUT fractions from
1129 one mouse, as well as the 2 SPQC samples. Samples corresponding to each experiment
1130 were concatenated and lyophilized to dryness.

1131 Samples were resuspended in 800 μ L 0.1% formic acid. 400 μ g was fractionated
1132 into 48 unique high pH reversed-phase fractions using pH 9.0 20 mM Ammonium formate
1133 as mobile phase A and neat acetonitrile as mobile phase B. The column used was a 2.1
1134 mm x 50 mm XBridge C18 (Waters) and fractionation was performed on an Agilent 1100

1135 HPLC with G1364C fraction collector. Throughout the method, the flow rate was 0.4
1136 mL/min and the column temperature was 55°C. The gradient method was set as follows:
1137 0 min, 3%B; 1 min, 7% B; 50 min, 50%B; 51 min, 90% B; 55 min, 90% B; 56 min, 3% B;
1138 70 min, 3% B. 48 fractions were collected in equal time segments from 0 to 52 minutes,
1139 then concatenated into 12 unique samples using every 12th fraction. For instance,
1140 fraction 1, 13, 25, and 37 were combined, fraction 2, 14, 26, and 38 were combined, etc.
1141 Fractions were frozen and lyophilized overnight. Samples were resuspended in 66 µL
1142 1%TFA/2% acetonitrile prior to LC-MS analysis.

1143 Quantitative LC/MS/MS was performed on 2 µL (1 µg) of each sample, using a
1144 nanoAcquity UPLC system (Waters) coupled to a Thermo Orbitrap Fusion Lumos high
1145 resolution accurate mass tandem mass spectrometer (Thermo) equipped with a FAIMS
1146 Pro ion-mobility device via a nanoelectrospray ionization source. Briefly, the sample was
1147 first trapped on a Symmetry C18 20 mm × 180 µm trapping column (5 µl/min at 99.9/0.1
1148 v/v water/acetonitrile), after which the analytical separation was performed using a 1.8
1149 µm Acquity HSS T3 C18 75 µm × 250 mm column (Waters) with a 90-min linear gradient
1150 of 5 to 30% acetonitrile with 0.1% formic acid at a flow rate of 400 nanoliters/minute
1151 (nL/min) with a column temperature of 55°C. Data collection on the Fusion Lumos mass
1152 spectrometer was performed for three different compensation voltages (CV: -40v, -60v, -
1153 80v). Within each CV, a data-dependent acquisition (DDA) mode of acquisition with a
1154 $r=120,000$ (@ m/z 200) full MS scan from m/z 375 – 1600 with a target AGC value of $4e^5$
1155 ions was performed. MS/MS scans were acquired in the Orbitrap at $r=50,000$ (@ m/z 200)
1156 from m/z 100 with a target AGC value of $1e^5$ and max fill time of 105 ms. The total cycle
1157 time for each CV was 1s, with total cycle times of 3 sec between like full MS scans. A 45s

dynamic exclusion was employed to increase depth of coverage. The total analysis cycle time for each sample injection was approximately 2 hours.

Following UPLC-MS/MS analyses, data were imported into Proteome Discoverer 2.4 (Thermo Scientific). The MS/MS data were searched against a SwissProt Mouse database (downloaded November 2019) plus additional common contaminant proteins, including yeast alcohol dehydrogenase (ADH), bovine casein, bovine serum albumin, as well as an equal number of reversed-sequence “decoys” for FDR determination. Mascot Distiller and Mascot Server (v 2.5, Matrix Sciences) were utilized to produce fragment ion spectra and to perform the database searches. Database search parameters included fixed modification on Cys (carbamidomethyl) and variable modification on Met (oxidation), Asn/Gln (deamidation), Lys (TMTPro) and peptide N-termini (TMTPro). Data were searched at 5 ppm precursor and 0.02 product mass accuracy with full trypsin enzyme rules. Reporter ion intensities were calculated using the Reporter Ions Quantifier algorithm in Proteome Discoverer. Percolator node in Proteome Discoverer was used to annotate the data at a maximum 1% protein FDR.

Mouse Behavioral Assays

Behavioral tests were performed on age-matched WT and homozygous SWIP^{P1019R} mutant littermates. Male and female mice were used in all experiments. Testing was performed at two time points: P42-55 days old as a young adult age, and 5.5 months old as mid-adulthood, so that we could compare disease progression in this mouse model to human patients (Ropers et al., 2011). The sequence of behavioral testing was: Y-maze (to measure working memory), object novelty recognition (to measure short-term and long-term object recognition memory), TreadScan (to assess gait), and steady-

speed rotarod (to assess motor control and strength) for 40-55 day old mice. Testing was performed over 1.5 weeks, interspersed with rest days for acclimation. This sequence was repeated with the same cohort at 5.5-6 months old, with three additional measures added to the end of testing: fear conditioning (to assess associative fear memory), a hearing test (to measure tone response), and a shock threshold test (to assess somatosensation). Of note, a separate, second cohort of mice was evaluated for fear conditioning, hearing, and shock threshold testing at adolescence. After each trial, equipment was cleaned with Labsan to remove residual odors. The experimenter was blinded to genotype for all behavioral analyses.

Y-maze

Working memory was evaluated by measuring spontaneous alternations in a 3-arm Y-maze under indirect illumination (80-90 lux). A mouse was placed in the center of the maze and allowed to freely explore all arms, each of which had different visual cues for spatial recognition. Trials were 5 min in length, with video data and analyses captured by EthoVision XT 11.0 software (Noldus Information Technology). Entry to an arm was defined as the mouse being >1 body length into a given arm. An alternation was defined as three successive entries into each of the different arms. Total % alternation was calculated as the total number of alternations/the total number of arm entries minus 2 x100.

Novel Object Recognition

One hour before testing, mice were individually exposed to the testing arena (a 48 x 22 x 18cm white opaque arena) for 10min under 80-100lux illumination without any objects. The test consisted of three phases: training (day 1), short-term memory test

(STM, day 1), and long-term memory test (LTM, day 2). For the training phase, two identical objects were placed 10 cm apart, against opposing walls of the arena. A mouse was placed in the center of the arena and given full access to explore both objects for 5 min and then returned to its home cage. For STM testing, one of the training objects remained (the now familiar object), and a novel object replaced one of the training objects (similar in size, different shape). The mouse was returned to the arena 30 minutes after the training task and allowed to explore freely for 5 mins. For LTM testing, the novel object was replaced with another object, and the familiar object remained unchanged. The LTM test was also 5 min in duration, conducted 24hr after the training task. Behavior was scored using Ethovision 11.0 XT software (Noldus) and analyzed by a blind observer. Object contact was defined as the mouse's nose within 1 cm of the object. We analyzed both number of nose contacts with each object and duration of contacts. Preference scores were calculated as $(\text{duration contact}_{\text{novel}} - \text{duration contact}_{\text{familiar}}) / \text{total duration contact}_{\text{novel} + \text{familiar}}$. Positive scores signified a preference for the novel object; whereas, negative scores denoted a preference for the familiar object, and scores approaching zero indicated no preference.

TreadScan

A TreadScan forced locomotion treadmill system (CleversSys Inc, Reston, Virginia) was used for gait recording and analysis. Each mouse was recorded walking on a transparent treadmill at 45 days old, and again at 5.5 months old. Mice were acclimated to the treadmill chamber for 1 minute before the start of recording to eliminate exploratory behavior confounding normal gait. Trials were 20 seconds in length, with mice walking at speeds between 13.83 and 16.53 cm/sec (P45 WT average 15.74 cm/s; P45 MUT

1227 average 15.80 cm/s; 5.5mo WT average 15.77 cm/s; 5.5mo MUT average 15.85 cm/s).
1228 A high-speed digital camera attached to the treadmill captured limb movement at a frame
1229 rate of 100 frames/second. We used TreadScan software (CleversSys) and
1230 representative WT and MUT videos to generate footprint templates, which were then used
1231 to identify individual paw profiles for each limb. Parameters such as stance time, swing
1232 time, step length, track width, and limb coupling were recorded for the entire 20 sec
1233 duration for each animal. Output gait tracking was verified manually by a blinded
1234 experimenter to ensure consistent limb tracking throughout the duration of each video.

1235 **Steady Speed Rotarod**

1236 A 5-lane rotarod (Med Associates, St. Albans, VT) was used for steady-speed
1237 motor analysis. The rod was run at a steady speed of 32rpm for four, 5-minute trials, with
1238 a 40-minute inter-trial interval. We recorded mouse latency to fall by infrared beam break,
1239 or manually for any mouse that completed two or more rotations on the rod without
1240 walking. Mice were randomized across lanes for each trial.

1241 **Fear Conditioning**

1242 Animals were examined in contextual and cued fear conditioning as described by
1243 Rodriguiz and Wetsel(Rodriguiz and Wetsel, 2006). Two separate cohorts of mice were
1244 used testing the two age groups. A three-day testing paradigm was used to assess
1245 memory: conditioning on day 1, context testing 24-hr post-conditioning on day 2, and cued
1246 tone testing 48hr post-conditioning on day 3. All testing was conducted in fear
1247 conditioning chambers (Med Associates). In the conditioning phase, mice were first
1248 acclimated to the test chamber for two minutes under ~100 lux illumination. Then a
1249 2900Hz, 80dB tone (conditioned stimulus, CS) played for 30 sec, which terminated with

1250 a paired 0.4mA, 2 sec scrambled foot shock (unconditioned stimulus, US). Mice were
1251 removed from the chamber and returned to their home cage 30 sec later. In the context
1252 testing phase, mice were placed in the same conditioning chamber and monitored for
1253 freezing behavior for a 5 min trial period, in the absence of the CS and US. For cued tone
1254 testing, the chambers were modified to different dimensions and shapes, contained
1255 different floor and wall textures, and lighting was adjusted to 50 lux. Mice acclimated to
1256 the chamber for 2 min, and then the CS was presented continuously for 3 min. Contextual
1257 and cued fear memory was assessed by freezing behavior, captured by automated video
1258 software (CleversSys).

1259 **Hearing Test**

1260 We tested mouse hearing using a startle platform (Med Associates) connected to
1261 Startle Pro Software in a sound-proof chamber. Mice were placed in a ventilated restraint
1262 cylinder connected to the startle response detection system to measure startle to each
1263 acoustic stimulus. After two minutes of acclimation, mice were assessed for an acoustic
1264 startle response to seven different tone frequencies, 2kHz, 3kHz, 4kHz, 8kHz, 12kHz,
1265 16kHz, and 20kHz that were randomly presented three times each at four different
1266 decibels, 80, 100, 105, and 110dB, for a total of 84 trials. A random inter-trial interval of
1267 15-60 seconds (average 30sec) was used to prevent anticipation of a stimulus. An
1268 animal's reaction to the tone was recorded as startle reactivity in the first 100msec of the
1269 stimulus presentation, which was transduced through the platform's load cell and
1270 expressed in arbitrary units (AU).

1271 **Startle Response (Somatosensation)**

Mouse somatosensation was tested by placing mice in a startle chamber (Med Associates) connected to Startle Pro Software. Mice were placed atop a multi-bar cradle within a ventilated plexiglass restraint cylinder, which allows for horizontal movement within the chamber, but not upright rearing. After two minutes of acclimation, each mouse was exposed to 10 different scrambled shock intensities, ranging from 0 to 0.6mA with randomized inter-trial intervals of 20-90 seconds. Each animal's startle reactivity during the first 100 msec of the shock was transduced through the platform's load cell and recorded as area under the curve (AUC) in arbitrary units (AU).

QUANTIFICATION AND STATISTICAL ANALYSIS

Experimental conditions, number of replicates, and statistical tests used are stated in each figure legend. Each experiment was replicated at least three times (or on at least 3 separate animals) to assure rigor and reproducibility. Both male and female age-matched mice were used for all experiments, with data pooled from both sexes. Data compilation and statistical analyses for all non-proteomic data were performed using GraphPad Prism (version 8, GraphPad Software, CA), using a significance level of $\alpha=0.05$. Prism provides exact p values unless $p<0.0001$. All data are reported as mean \pm SEM. Each data set was tested for normal distribution using a D'Agostino-Person normality test to determine whether parametric (unpaired Student's t-test, one-way ANOVA, two-way ANOVA) or non-parametric (Mann-Whitney, Kruskal-Wallis) tests should be used. Parametric assumptions were confirmed with the Shapiro-Wilk test (normality) and Levine's test (error variance homogeneity) for ANOVA with repeated

measures testing. The analysis of iBioID and TMT proteomics data are described below.
All proteomic data and analysis scripts are available online (see Resource Availability).

Imaris 3D reconstruction

For EEA1+ and CathepsinD+ puncta analyses, coverslips were imaged on a Zeiss LSM 710 confocal microscope. Images were sampled at a resolution of 1024 x 1024 pixels with a dwell time of 0.45µsec using a 63x/1.4 oil immersion objective, 2.0 times digital zoom, and a z-step size of 0.37 µm. Images were saved as “.ism” formatted files, and quantification was performed on a POGO Velocity workstation in the Duke Light Microscopy Core Facility using Imaris 9.2.0 software (Bitplane, South Windsor, CT). For analyses, we first used the “surface” tool to make a solid fill surface of the MAP2-stained neuronal soma and dendrites, with the background subtraction option enabled. We selected a threshold that demarcated the neuron structure accurately while excluding background. For EEA1 puncta analyses, a 600 x 800 µm selection box was placed around the soma in each image and surfaces were created for EEA1 puncta within the selection box. Similarly, for CathepsinD puncta analyses, a 600 x 600 µm selection box was placed around the soma(s) in each image for surface creation. The same threshold settings were used across all images, and individual surface data from each soma were exported for aggregate analyses. The experimenter was blinded to sample conditions for both image acquisition and analysis.

Cleaved Caspase-3 Image Analysis

Z-stack images were acquired on a Zeiss 710 LSM confocal microscope. Images were sampled at a resolution of 1024 x 1024 pixels with a dwell time of 1.58µsec, using a 63x/1.4 oil immersion objective (for cortex and striatum) or 20x/0.8 dry objective

(cerebellum), a 1.0 times digital zoom, and a z-step size of 0.67 μm . Images were saved as “.ism” formatted files, and then converted into maximum intensity projections (MIP) using Zen 2.3 SP1 software. Quantification of CC3 colocalization with DAPI was performed on the MIPs using the Particle Analyzer function in FIJI ImageJ software. The experimenter was blind to sample conditions for both image acquisition and analysis.

iBioID Quantitative Analysis

Following UPLC-MS/MS analyses, data was imported into Proteome Discoverer 2.2 (Thermo Scientific Inc.), and aligned based on the accurate mass and retention time of detected ions (“features”) using Minora Feature Detector algorithm in Proteome Discoverer. Relative peptide abundance was calculated based on area-under-the-curve (AUC) of the selected ion chromatograms of the aligned features across all runs. The MS/MS data was searched against the SwissProt *Mus musculus* database (downloaded in April 2018) with additional proteins, including yeast ADH1, bovine serum albumin, as well as an equal number of reversed-sequence “decoys” for false discovery rate (FDR) determination. Mascot Distiller and Mascot Server (v 2.5, Matrix Sciences) were utilized to produce fragment ion spectra and to perform the database searches. Database search parameters included fixed modification on Cys (carbamidomethyl), variable modifications on Meth (oxidation) and Asn and Gln (deamidation), and were searched at 5 ppm precursor and 0.02 Da product mass accuracy with full trypsin enzymatic rules. Peptide Validator and Protein FDR Validator nodes in Proteome Discoverer were used to annotate the data at a maximum 1% protein FDR.

Protein intensities were exported from Proteome Discoverer and processed using custom R scripts. Carboxylases and keratins, as well as 315 mitochondrial proteins(Calvo

et al., 2016), were removed from the identified proteins as known contaminants. Next, we performed sample loading normalization to account for technical variation between the 9 individual MS runs. This is done by multiplying intensities from each MS run by a scaling factor, such that the average of all total run intensities are equal. As QC samples were created by pooling equivalent aliquots of peptides from each biological replicate, the average of all biological replicates should be equal to the average of all technical SPQC replicates. We performed sample pool normalization to SPQC samples to standardize protein measurements across all samples and correct for batch effects between MS analyses. Sample pool normalization adjusts the protein-wise mean of all biological replicates to be equal to the mean of all SPQC replicates. Finally, proteins that were identified by a single peptide, and/or identified in less than 50% of samples were removed. Any remaining missing values were inferred to be missing not at random due to the left shifted distribution of proteins with missing values and imputed using the k-nearest neighbors algorithm using the `impute.knn` function in the R package `impute` (`impute::impute.knn`). Normalized protein data was analyzed using `edgeR`, an R package for the analysis of differential expression/abundance that models count data using a binomial distribution methodology. Differential enrichment of proteins in the WASH1-BioID2 pull-down relative to the solubleBioID2 control pull-down were evaluated with an exact test as implemented by the `edgeR::exactTest` function. To consider a protein enriched in the WASH interactome, we required that a protein exhibit a fold change greater than 3 over the negative control with an exact test Benjamini Hochberg adjusted p-value (FDR) less than 0.1. With these criteria, 174 proteins were identified as WASH1

interactome proteins. Raw peptide and final normalized protein data as well as the statistical results can be found in Supplementary Table 1.

Proteins that function together often interact directly. We compiled experimentally-determined protein-protein interactions (PPIs) among the WASH1 interactome from the HitPredict database(López et al., 2015) using a custom R package, getPPIs, (available online at [twesleyb/getPPIs](https://twesleyb.github.io/getPPIs/)). We report PPIs among the WASH1 interactome in Supplemental Table 1.

Bioinformatic GO analysis was conducted by manual annotation of identified proteins and confirmed with Metascape analysis(Zhou et al., 2019) of WASH1-BioID2 enriched proteins using the 2,311 proteins identified in the mass spec analysis as background.

Raw peptide intensities were exported from Proteome Discover for downstream analysis and processing in R. Following database searching, protein scoring using the Protein FDR Validator algorithm, and removal of contaminant species, the dataset retained 86,551 peptides corresponding to the identification of 7,488 unique proteins. These data, as well as statistical results can be found in Supplemental Table 2.

TMT Proteomics Quantitative Analysis

Peptide level data from the spatial proteomics analysis of SWIP^{P1019R} MUT and MUT brain were exported from Proteome Discoverer (version 2.4) and analyzed using custom R and Python scripts. Peptides from contaminant and non-mouse proteins were removed. First, we performed sample loading normalization, normalizing the total ion intensity for each TMT channel within an experiment to be equal. Sample loading normalization corrects for small differences in the amount of sample analyzed and

labeling reaction efficiency differences between individual TMT channels within an experiment.

We found that in each TMT experiment there were a small number of missing values (mean percent missing = 1.6 +/- 0.17%). Missing values were inferred to be missing at random based on the overlapping distributions of peptides with missing values and peptides without missing values. We imputed these missing values using the k-nearest neighbor algorithm (`impute::impute.knn`). Missing values for SPQC samples were not imputed. Peptides with any missing SPQC data were removed.

Following sample loading normalization, SPQC replicates within each experiment should yield identical measurements. As peptides with irreproducible QC measurements are unlikely to be quantitatively robust, and their inclusion may bias downstream processing (see IRS normalization below), we sought to remove them. To assess intra-batch variability, we utilized the method described by Ping et al., 2019(Ping et al., 2018). Briefly, peptides were binned into 5 groups based on the average intensity of the two SPQC replicates. For each pair of SPQC measurements, the log ratio of SPQC intensities was calculated. To identify outlier QC peptides, we plotted the distribution of these log ratios for each bin. Peptides with ratios that were more than four standard deviations away from the mean of its intensity bin were considered outliers and removed (total number of SPQC outlier peptides removed = 474).

Proteins were summarized as the sum of all unique peptide intensities corresponding to a unique UniProtKB Accession identifier, and sample loading normalization was performed across all three experiments to account for inter-experimental technical variability. In a TMT experiment, the peptides selected for MS2

fragmentation for any given protein is partially random, especially at lower signal-to-noise peptides. This stochasticity means that proteins are typically quantified by different peptides in each experiment. Thus, although SPQC samples should yield identical protein measurements in each of the three experiments (as it is the same sample analyzed in each experiment), the observed protein measurements exhibit variability due to their quantification by different peptides. To account for this protein-level bias, we utilized the internal reference scaling (IRS) approach described by Plubell et al., 2017(Plubell et al., 2017). IRS normalization scales the protein-wise geometric average of all SPQC measurements across all experiments to be equal, and simultaneously adjusts biological replicates. In brief, each protein is multiplied by a scaling factor which adjusts its intra-experimental SPQC values to be equal to the geometric mean of all SPQC values for the three experiments. This normalization step effectively standardizes protein measurements between different mass spectrometry experiments.

The final normalization step was to perform sample pool normalization using SPQC samples as a reference. This normalization step, sometimes referred to as global internal standard normalization, accounts for batch effects between experiments, and reflects the fact that after technical normalization, the mean of biological replicates should be equal to the mean of SPQC replicates.

Before assessing protein differential abundance, we removed irreproducible proteins. This included proteins that were quantified in less than 50% of all samples, proteins that were identified by a single peptide, and proteins that had missing SPQC values. Across all 42 biological replicates, we observed that a small number of proteins had potential outlier measurements that were either several orders of magnitude greater

or less than the mean of its replicates. In order to identify and remove these proteins, we assessed the reproducibility of protein measurements within a fraction in the same manner used to identify and filter SPQC outlier peptides. A small number of proteins were identified as outliers if the average log ratio of their 3 technical replicates was more than 4 standard deviations away from the mean of its intensity bin (n=349). In total, we retained 5,897 of the original 7,488 proteins in the final dataset.

Differential protein abundance was assessed using the final normalized protein data for intrafraction comparisons between WT and MUT groups using a general linear model as implemented by the `edgeR::glmQLFit` and `edgeR::glmQLFTest` functions (MD et al., 2009). Although this approach was originally devised for analysis of single-cell RNA-sequencing data, this approach is also appropriate for proteomics count data which is over-dispersed, negative binomially distributed, and often only includes a small number of replicates (for an example of edgeR's application to proteomics see Plubell *et al.*, 2017 (Plubell et al., 2017)) (McCarthy et al., 2012; MD et al., 2009). For intrafraction comparisons, P-values were corrected using the Benjamini Hochberg procedure within edgeR. An FDR threshold of 0.1 was set for significance for intrafraction comparisons.

We utilized edgeR's flexible GLM framework to test the hypothesis that the abundance of proteins in the WT group was significantly different from that in the MUT group irrespective of fraction differences (Supplementary Table 2). For WT vs. MUT contrasts, we considered proteins with an FDR < 0.05 significant (n=687). For plotting, we adjusted normalized protein abundances for fraction differences by fitting the data with an additive linear model with fraction as a blocking factor, as implemented by the `removeBatchEffect` algorithm from the R limma package (Ritchie et al., 2015).

To construct a protein covariation graph, we assessed the pairwise covariation (correlation) between all 5,897 proteins using the biweight midcorrelation (WGCNA::bicor) statistic, a robust alternative to Pearson's correlation. The resulting complete, signed, weighted, and symmetric adjacency matrix was then re-weighted using Network Enhancement. Network enhancement removes noise from the graph, and facilitates downstream community detection(Wang et al., 2018).

The enhanced adjacency matrix was clustered using the Leiden algorithm(Traag et al., 2019), a recent extension and improvement of the well-known Louvain algorithm(Mucha et al., 2010). The Leiden algorithm functions to optimize the partition of a graph into modules by maximizing a quality statistic. We utilized the 'Surprise' quality statistic(Traag et al., 2015) to identify optimal partitions of the protein covariation graph. To facilitate biological inferences drawn from the network's organization, we recursively split 27 modules that contained more than 100 nodes and removed modules that were smaller than 5 proteins. Initial clustering of the network resulted in the identification of 324 modules.

To reduce the likelihood of identifying false positive modules, we enforced module quality using a permutation procedure (NetRep::modulePreservation)(Ritchie et al., 2016) and removed modules with any insignificant permutation statistics (Bonferroni P-Adjust > 0.05). The following statistics were used to enforce module quality: 'avg.weight' (average edge weight), 'avg.cor' (average bicor correlation R^2), and 'avg.contrib' (quantifies how similar an individual protein's abundance profile is to the summary of its module). Proteins which were assigned to modules with insignificant module quality statistics were not considered clustered as the observed quality of their module does not differ from random.

After filtering, approximately 85% of all proteins were assigned a cluster. The median percent variance explained by the first principle component of a module (a measure of module cohesiveness) was high (59.8%). After removal of low-quality modules, the analysis retained 255 distinct modules of proteins that strongly covaried together (Supplemental Table 3).

To evaluate modules that were changing between WT and MUT genotypes, we extended the GLM framework to test for protein differential abundance. Modules were summarized as the sum of their proteins and fit with a GLM, with fraction as a blocking factor. In this statistical design, we were interested in the average effect of genotype on all proteins in a module. For plotting, module abundance was adjusted for fraction differences using the `removeBatchEffect` function (package: `limma`). We utilized the Bonferroni method to adjust P-values for 255 module level comparisons and considered modules with an adjusted P-value less than 0.05 were considered significant (n=37).

SUPPLEMENTARY FILES

- **Supplementary Table 1:** WASH iBioID raw and normalized data and the corresponding statistical results.
- **Supplementary Table 2:** SWIP^{P1019R} TMT raw and normalized proteomics data and the corresponding statistical results.
- **Supplementary Table 3:** Module-level data and statistical results from network analysis of SWIP^{P109R} proteomics.
- **Supplementary Table 4:** DNA oligonucleotides used in this study.

SUPPLEMENTAL FIGURE TITLES AND LEGENDS

Figure S1. Overexpression of SWIP^{P1019R} decreases WASH complex binding in cultured cells; related to Figure 2

(A) Schematic showing overexpression of WT or MUT SWIP^{P1019R} in HEK293T cells followed by immunoprecipitation.

(B) Western blots of input (5%, left) and immunoprecipitated (IP, right) protein. Two samples per condition were run on two separate gels, n=4 separate experiments.

(C) Quantification of B normalized to WT. Strumpellin (WT 100.0 ± 6.8%, MUT 54.8 ± 8.0%, $t_{5.9}=4.290$, $p=0.0054$), WASH1 (WT 100.0 ± 7.3%, MUT 41.4 ± 4.4%, $t_{4.9}=6.902$, $p=0.0011$), HA (WT 100.0 ± 4.1%, MUT 107.8 ± 4.1%, $t_{6.0}=1.344$, $p=0.2275$). Data reported as mean ± SEM, error bars are SEM. ** $p<0.01$, two-tailed t-tests.

Figure S2. SWIP^{P1019R} MUT brain displays significant alterations in protein abundance compared to WT; related to Figures 2 and 3

(A) Interactome of altered proteins. Nodes reflect protein name, light gray lines delineate proteins identified as different in MUT compared to WT (Δ WASHC4 in center), dark dashed lines indicate known protein-protein interactions from HitPredict database (López et al., 2015). Color reflects cellular function seen in B. Nodes with red borders delineate WASH complex proteins.

(B) Cellular function of proteins in A, as reflected in published literature. % reflects the percentage of proteins in a category out of the total 85 altered proteins.

Clustergrams of:

1522 (C) Proteins with increased (red) or decreased (blue) abundance in MUT brains compared
1523 to WT.

1524 (D) Known protein components of the WASH complex and their previously reported
1525 interactors.

1526 (E) Proteins with lysosomal function.

1527 (F) Proteins identified in the WASH1-BioID2 proteome (Figure 1).

1528 (G) Proteins with links to intellectual disability (I.D.) or neurodegeneration (Degen.).

1529

1530 **Figure S3. Multiple protein networks display significant alterations in MUT brain**
1531 **compared to WT; related to Figures 2 and 3**

1532 (A) PCA of all proteins identified by TMT proteomics. Each protein is color-coded
1533 according to its module assignment. PC = principal component, PVE = percent variance
1534 explained.

1535 Network graphs of:

1536 (B) Module 14 (M14) containing endosomal proteins.

1537 (C) Module 83 (M83) containing endoplasmic reticulum (ER) stress response proteins.

1538 (D-E) Modules 143 and 35 (M143, M35, respectively) containing synaptic proteins.

1539 Network attributes graphs in B-E: Node size denotes its weighted degree centrality
1540 (~importance in module), colored node indicates altered abundance in MUT brain relative
1541 to WT, black node border denotes proteins identified in the WASH1-BioID proteome
1542 (Figure 1), black edges indicate known protein-protein interactions, and grey-red edges
1543 denote the relative strength of protein covariation within a module (gray = weak, red =
1544 strong).

(F) Summary box plots of module abundance for all seven brain fractions analyzed presented as \log_2 (adjusted module intensities). All modules display significant differences between WT and MUT groups (M14: WT 13.94 ± 0.002 MUT 13.82 ± 0.002 , $p=0.0207$; M83: WT 14.46 ± 0.004 , MUT 14.70 ± 0.003 , $p=0.0088$; M143: WT 14.03 ± 0.002 , MUT 13.92 ± 0.002 , $p=0.0148$; M35: WT 14.19 ± 0.001 , MUT 14.10 ± 0.001 , $p=0.0030$).

(G) Normalized protein abundance for all proteins in each module across the seven subcellular fractions analyzed. Plots correspond to modules seen in F, and data colors reflect genotypes in F. Number of proteins in each module are indicated at the top of each graph. Data reported as mean \pm SEM, error bars are SEM. * $p<0.05$, ** $p<0.01$, *** $p<0.001$, GLM (model: 0 ~ Fraction + Module) and empirical Bayes quasi-likelihood F-test with Bonferroni correction (G).

Figure S4. There is no significant difference in striatal or cerebellar cell death between WT and MUT mice; related to Figure 5

(A) Representative image of adolescent (P42) WT striatum stained with cleaved caspase-3 (CC3, green).

(B) Representative image of adolescent (P42) MUT striatum stained with cleaved caspase-3 (CC3, green).

(C and D) DAPI co-stained images of A and B, respectively.

(E) Anatomical representation of mouse brain with striatum highlighted in red, adapted from Allen Brain Atlas (Oh et al., 2014).

(F) Representative image of adult (10 mo) WT striatum stained with CC3 (green).

(G) Representative image of adult (10 mo) MUT striatum stained with CC3 (green).

1568 (H and I) DAPI co-stained images of F and G, respectively. Scale bars for A-I are 15 μ m.
1569 (J) Graph depicting the % of DAPI+ nuclei that are positive for CC3 per image. No
1570 difference is seen between genotypes at either age (P42 WT $3.70 \pm 0.99\%$, P42 MUT
1571 $1.95 \pm 0.49\%$, 10mo WT $16.77 \pm 2.09\%$, 10mo MUT $24.86 \pm 2.17\%$, $H=61.87$, $p<0.0001$).
1572 (K) Representative image of adolescent (P42) WT cerebellum stained with cleaved
1573 caspase-3 (CC3, green).
1574 (L) Representative image of adolescent (P42) MUT cerebellum stained with cleaved
1575 caspase-3 (CC3, green).
1576 (M and N) DAPI and Calbindin co-stained images of K and L, respectively.
1577 (O) Anatomical representation of mouse cerebellum, adapted from Allen Brain Atlas (Oh
1578 et al., 2014). Red region highlights area used for imaging.
1579 (P) Representative image of adult (10 mo) WT cerebellum stained with CC3 (green).
1580 (Q) Representative image of adult (10 mo) MUT cerebellum stained with CC3 (green). No
1581 significant CC3 staining is observed at either age.
1582 (R and S) DAPI and Calbindin co-stained images of P and Q, respectively. Scale bars for
1583 K-S are 50 μ m.
1584 (T) Graph depicting the number of Calbindin+ somas per image, a marker for Purkinje
1585 cells. No difference is seen between genotypes at either age (P42 WT 20.50 ± 0.53 , P42
1586 MUT 20.67 ± 0.59 , 10mo WT 21.42 ± 0.85 , 10mo MUT 22.63 ± 0.74 , $H=4.891$, $p=0.1799$).
1587 Data obtained from four animals per condition, and reported as mean \pm standard error of
1588 the mean (SEM), with error bars as SEM. Kruskal-Wallis test (J and T).
1589

Figure S5. SWIP^{P1019R} mutant mice do not display deficits in spatial working memory or novel object recognition; related to Figure 6

(A) Y-maze paradigm. Mice were placed in the center of the maze and allowed to explore all three arms freely for five minutes. Each arm had distinct visual cues.

(B) Graph depicting the percent alternations achieved for each mouse at adolescence (WT $51.48 \pm 2.47\%$, MUT $50.81 \pm 2.19\%$, $t_{24}=0.2036$, $p=0.8404$).

(C) Graph depicting the percent alternations achieved for each mouse at adulthood (WT $56.12 \pm 1.53\%$, MUT $57.93 \pm 2.56\%$, $t_{18}=0.6074$, $p=0.5511$).

(D) Graphs of the number of direct revisits mice made to the arm they just explored reveal no difference between genotypes at adolescence (WT 2.64 ± 0.50 , MUT 2.42 ± 0.42 , $t_{24}=0.3482$, $p=0.7308$).

(E) Graphs of the number of direct revisits mice made at adulthood (WT 1.36 ± 0.33 , MUT 1.42 ± 0.36 , $t_{24}=0.1231$, $p=0.9031$).

(F) Similar to D-E, there were no differences in the number of indirect revisits (ex: arm A \rightarrow arm B \rightarrow arm A) between genotypes at adolescence (WT 10.21 ± 1.03 , MUT 12.00 ± 1.48 , $U=69$, $p=0.4515$).

(G) Indirect revisits at adulthood (WT 12.86 ± 1.26 , MUT 15.17 ± 1.43 , $t_{23}=1.211$, $p=0.2380$).

(H) There were no significant differences in total distance travelled at adolescence, suggesting that motor function did not affect Y-maze performance (WT 2401 ± 98.9 cm, MUT 2406 ± 121.0 cm, $t_{22}=0.03281$, $p=0.9741$).

(I) No difference in distance travelled at adulthood (WT 2761 ± 111.6 cm, MUT 3124 ± 191.1 cm, $t_{18}=1.638$, $p=0.1189$).

1613 (J) Novel object task. Mice first performed a 5-minute trial in which they were placed in
 1614 an arena with two identical objects and allowed to explore freely, while their behavior was
 1615 tracked with video software. Mice were returned to their home cage, and then re-
 1616 introduced to the arena a half an hour later, where one of the objects had been replaced
 1617 with a novel object, and their behavior was again tracked. Twenty-four hours later the
 1618 same test was performed, but the novel object was replaced with another new object.

1619 (K) Graph depicting adolescent animals' preference for the novel object during the three
 1620 phases of the task, training (Train), short-term memory (STM), and long-term memory
 1621 (LTM). No significant difference in object preference is seen between genotypes for any
 1622 phase (Train WT -0.092 ± 0.065 , Train MUT -0.123 ± 0.072 , STM WT 0.488 ± 0.076 , STM
 1623 MUT 0.315 ± 0.094 , LTM WT 0.479 ± 0.046 , LTM MUT 0.373 ± 0.076 , $F_{1,22}=1.840$,
 1624 $p=0.1887$).

1625 (L) Graph depicting adult animals' preference for the novel object. (Train WT $-0.066 \pm$
 1626 0.053 , Train MUT -0.130 ± 0.089 , STM WT 0.416 ± 0.060 , STM MUT 0.274 ± 0.096 , LTM
 1627 WT 0.316 ± 0.059 , LTM MUT 0.306 ± 0.050 , $F_{1,22}=0.9735$, $p=0.3345$).

1628 (M) Graph depicting the total amount of time (in seconds) adolescent animals spent
 1629 exploring both objects in each phase of the task. No significant difference in exploration
 1630 time was observed across genotypes, suggesting that genotype does not hinder object
 1631 exploration (Train P44 WT 38.10 ± 2.45 s, Train P44 MUT 50.04 ± 5.44 s, STM P44 WT
 1632 55.02 ± 4.31 s, STM P44 MUT 63.60 ± 4.50 s, LTM P45 WT 46.75 ± 3.34 s, LTM P45
 1633 MUT 68.08 ± 7.54 s, $F_{1,22}=7.373$, $p=0.0126$).

1634 (N) Graph depicting the total time adult animals spent exploring both objects in each the
 1635 task (Train WT 39.10 ± 3.62 s, Train MUT 51.15 ± 4.91 s, STM WT 44.75 ± 4.87 s, STM

MUT 49.56 ± 6.16 s, LTM WT 39.02 ± 5.29 s, LTM MUT 55.15 ± 7.34 s, $F_{1,22}=2.936$, $p=0.1007$). For all Y-maze measures, WT $n=14$, MUT $n=12$. For all novel object measures, WT $n=13$, MUT $n=11$. Data reported as mean \pm SEM, error bars are SEM. Two-tailed t-tests or Mann-Whitney U tests (B-I), two-way ANOVAs (K-N).

Figure S6. SWIP^{P1019R} mutant mice do not have significant deficits in contextual fear memory recall, auditory perception, or tactile sensation; related to Figure 6

(A) Experimental fear conditioning scheme. After acclimation, mice received a mild aversive 0.4mA footshock paired with a 2900Hz tone in a conditioning chamber. 24 hours later, the mice were placed back in the same chamber to assess freezing behavior (without footshock or tone).

(B) Experimental startle response setup used to assess hearing and somatosensation. Mice were placed in a plexiglass tube atop a load cell that measured startle movements in response to stimuli.

(C) Line graphs of WT and MUT freezing response during the contextual memory recall task. Data represented as average freezing per genotype in 30 second time bins. The task was performed with two different cohorts for the different ages, P42 (top) and 6.5mo (bottom). Top: no significant difference in freezing at P42 (Two-way repeated measure ANOVA, Genotype effect, $F_{1,18}=0.088$, $p=0.7698$. Sidak's post-hoc analysis, 30 s $p=0.8388$, 60 s $p=0.9990$, 90 s $p=0.9964$, 120 s $p=0.3281$), Bottom: no significant difference in freezing at 6.5mo (Two-way repeated measure ANOVA, Genotype effect, $F_{1,22}= 3.723$, $p=0.0667$. Sidak's post-hoc analysis, 30 s $p=0.8977$, 60 s $p=0.1636$, 90 s $p=0.9979$, 120 s $p=0.0037$).

(D) Graphs showing the average total freezing time per animal during context exposure. Top: no significant difference is seen between WT and MUT mice at P42 (WT $34.01 \pm 6.32\%$, MUT $36.99 \pm 7.81\%$, $t_{17}=2.985$, $p=0.7699$). Bottom: no significant difference is seen between genotypes at 6.5mo (WT $43.94 \pm 6.00\%$, MUT $28.73 \pm 4.80\%$, $t_{21.6}=1.980$, $p=0.0606$).

(E) Graphs of individual animals' startle response to a 2900Hz tone played at 80dB. MUT mice were not significantly more reactive to the tone than WT at P50 (WT 25.96 ± 4.95 , MUT 40.68 ± 5.05 , $U=35$, $p=0.2799$), or at 6.5mo (WT 14.07 ± 3.27 , MUT 14.85 ± 1.49 , $U=47$, $p=0.2768$).

(F) Graphs of individual animals' startle response to a 0.4mA footshock. No significant difference observed between genotypes at either age (P55 WT 1527 ± 215.7 , P55 MUT 1996 ± 51.0 , $U=28.50$, $p=0.0542$; 6.5mo WT 1545 ± 179.5 , 6.5mo MUT 1817 ± 119.1 , $U=47$, $p=0.2360$). Startle response reported in arbitrary units (A.U.). For all adolescent measures: WT $n=10$, MUT $n=10$. For adult freezing measures: WT $n=13$, MUT $n=11$. For adult startle responses: WT $n=13$, MUT $n=10$. Data reported as mean \pm SEM, error bars are SEM. * $p<0.05$, ** $p<0.01$, two-way repeated measure ANOVAs (C), two-tailed t-tests (D), and Mann-Whitney U tests (E-F).

Figure S7. Progressive gait changes in SWIP^{P1019R} mutant mice are not restricted to rear limbs; related to Figure 7

(A) Graph of average swing time per stride for front limbs. At P45 (top), there is no significant difference in front swing time (WT 177.5 ± 10.9 ms, MUT 175.3 ± 8.7 ms,

1681 $t_{24}=0.1569$, $p=0.8766$). At 5.5 mo (bottom), MUT mice take significantly longer to swing
1682 their forelimbs (WT 178.6 ± 6.2 ms, MUT 206.3 ± 7.2 ms, $t_{21.4}=2.927$, $p=0.0079$).

1683 (B) Graph of average stride length for front limbs. At P45 (top), there is no difference in
1684 WT and MUT stride length (WT 57.0 ± 0.9 mm, MUT 59.2 ± 1.4 mm, $U=68$, $p=0.2800$).
1685 At 5.5mo (bottom), MUT mice take significantly longer strides with their forelimbs (WT
1686 60.0 ± 0.6 mm, MUT 63.7 ± 0.9 mm, $t_{17}=3.545$, $p=0.0024$).

1687 (C) Graph of average homolateral coupling, the fraction of a reference foot's stride when
1688 its ipsilateral foot starts its stride. At P45, there is no significant difference in homolateral
1689 coupling (WT 0.48 ± 0.005 , MUT 0.48 ± 0.004 , $U=90$, $p=0.9713$), but at 5.5mo, MUT mice
1690 display decreased homolateral coupling (WT 0.48 ± 0.002 , MUT 0.45 ± 0.005 , $t_{13.5}=3.469$,
1691 $p=0.0039$).

1692 (D) Graph of average track width between front limbs. At P45 (top), there is a significantly
1693 narrower front track width in MUT compared to WT (WT 16.99 ± 0.15 mm, MUT $15.12 \pm$
1694 0.33 mm, $t_{17}=5.192$, $p<0.0001$). This difference persists into adulthood at 5.5mo (WT
1695 19.36 ± 0.23 mm, MUT 16.74 ± 0.46 mm, $t_{15}=5.055$, $p=0.0001$).

1696 (E) Graph of average track width between rear limbs. At P45 (top), there is no difference
1697 in WT and MUT rear track widths (WT 29.58 ± 0.51 mm, MUT 28.77 ± 0.36 mm, $t_{23}=1.292$,
1698 $p=0.2091$). At 5.5mo (bottom), mutants display significantly narrower rear track widths
1699 (WT 32.59 ± 0.34 mm, MUT 30.01 ± 0.46 mm, $t_{19.4}=4.502$, $p=0.0002$). For P45 measures:
1700 WT $n=14$, MUT $n=13$; for 5.5mo measures: WT $n=14$, MUT $n=11$. Data reported as mean
1701 \pm SEM, error bars are SEM. ** $p<0.01$, *** $p<0.001$, **** $p<0.0001$, two-tailed t-tests or
1702 Mann-Whitney U tests.

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