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## Amyloid accumulation drives proteome-wide alterations in mouse models of Alzheimer's disease like pathology

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### SUMMARY

Amyloid beta (A $\beta$ ) peptides impair multiple cellular pathways and play a causative role in Alzheimer's disease (AD) pathology, but how the brain proteome is remodeled by this process is unknown. To identify protein networks associated with AD-like pathology, we performed global quantitative proteomic analysis in three mouse models at young and old ages. Our analysis

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### SUPPLEMENTARY INFORMATION

Supplemental Information includes Supplemental Experimental Procedures and six figures and six tables can be found with this article online at:

### AUTHOR CONTRIBUTIONS

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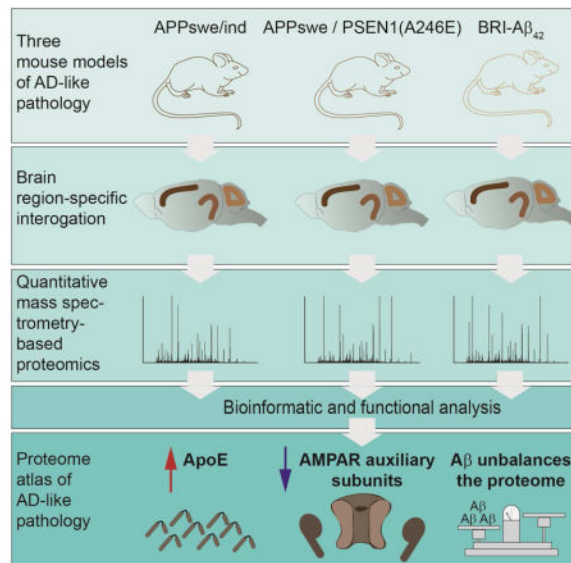
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revealed a robust increase in Apolipoprotein E (ApoE) levels in nearly all brain regions with increased A $\beta$  levels. Taken together with prior findings on ApoE driving A $\beta$  accumulation, this analysis points to a pathological dysregulation of the ApoE-A $\beta$  axis. We also found dysregulation of protein networks involved in excitatory synaptic transmission. Analysis of the AMPA receptor (AMPA) complex revealed specific loss of TARP $\gamma$ -2, a key AMPAR trafficking protein. Expression of TARP $\gamma$ -2 in hAPP transgenic mice restored of AMPA currents. This proteomic database represents a resource for the identification of protein alterations responsible for AD.

## Graphical Abstract



## Keywords

AD; proteomics; mass spectrometry; synapses; proteostasis; WGCNA; amyloid beta; ApoE; AMPAR

## INTRODUCTION

AD is a progressive brain disorder that is the leading cause of dementia in adults, resulting in impaired memory and cognition. Pathological hallmarks include amyloid plaques, neurofibrillary tangles, astrogliosis, and changes in brain vasculature, which culminate in neurodegeneration. Abundant evidence shows AD brain pathology is complex and represents feedback and feed forward responses in multiple cell types (De Strooper and Karran, 2016). Many mechanisms have been proposed to explain AD, but a unifying and universally accepted description has yet to be achieved. There is near unanimous agreement that A $\beta$  peptides generated by proteolytic processing of amyloid precursor protein (APP) represent a key toxic species. The “amyloid cascade hypothesis” has provided a useful theoretical framework for the basis of AD (Selkoe and Hardy, 2016), yet current evidence suggests a more complicated pathological etiology (Herrup, 2015). One major challenge impeding our understanding of AD pathology is that the molecular perturbations vary based on the stage

of disease and brain region. Furthermore, even individuals with mutations in APP or Presenilin-1 (PS1) take decades to develop AD, suggesting that long incubation periods or aging is required. Several molecular pathways including impaired synaptic transmission, hampered protein degradation dynamics, defective mitochondrial function, and inflammatory responses all play important roles in AD pathogenesis (Akiyama et al., 2000; Lin and Beal, 2006; Mori et al., 1987; Sheng et al., 2012).

Dynamic changes in protein abundances represent key cellular responses to cope with cellular insults. However, broad toxicity, such as accumulation of A $\beta$  peptides, likely affects many proteins and pathways. We hypothesized this must ultimately result in proteome remodeling (Powers et al., 2009). For example, proteins physically associated with A $\beta$  plaques accumulate while proteins localized to synapses have reduced levels in AD (Lassmann et al., 1992). To understand the effects of A $\beta$  peptides on the brain proteome, we performed quantitative proteomic analysis of frontal cortex (FC), hippocampal (HIP), and cerebellar (CB) extracts from transgenic mice expressing mutated hAPP or hAPP/PS1 (Borchelt et al., 1996; Mucke et al., 2000). While these models have provided insight into AD pathology, they are limited by over-expression of APP fragments outside A $\beta$  in non-relevant cells (Saito et al., 2014). We also performed proteomic analysis of BRI-A $\beta$ 42 mouse brain extracts over-expressing A $\beta$ 42 without additional APP fragments (McGowan et al., 2005). To temporally resolve changes in protein abundance, we performed our analysis at ages before and after plaques form to find early and potentially reversible alterations.

We report an AD proteomic resource of 18,882 quantified proteins (10,288 genes) from mouse model brain region extracts. To delineate proximal effects of A $\beta$  on the brain proteome, we determined the level of A $\beta$  and amyloid plaques and performed our analyses at a- and symptomatic time points, based on previous reports (Dodart et al., 1999; Lalonde et al., 2005). We observed an age dependent increase in proteome remodeling, region specific patterns, and found many proteins genetically linked to late onset AD (LOAD), such as ApoE, as significantly altered. ApoE plays a key role in regulating A $\beta$  metabolism and is co-deposited in plaques (Liu et al., 2013; Namba et al., 1991). ApoE also seems to play a role in A $\beta$  aggregation (Bales et al., 1997). However, how ApoE levels are altered in AD remains unclear (Gupta et al., 2011; Hesse et al., 2000; Kuo et al., 2000; Liu et al., 2013). We found ApoE significantly increased in all three models and only in those datasets with increased A $\beta$  levels. Overall; we identified a small panel of proteins significantly altered in multiple models. Co-expression analysis revealed two distinct modules (ME) involved with synaptic transmission in HIP and two MEs related to mitochondria in FC. Biochemical purification of AMPARs, which represented core module proteins, showed a specific loss of TARP $\gamma$ -2 (known as stargazin). To assess if our proteomics can lead to important conclusions, we tested and found that correcting TARP $\gamma$ -2 levels can restore glutamatergic synaptic transmission *in vivo*.

## RESULTS

### Remodeled proteomes in hAPP and hAPP/PS1 brains

To determine how A $\beta$  peptide toxicity affects the brain proteome, we developed a quantitative proteomic strategy using nitrogen-15 ( $^{15}\text{N}$ ) labeled mice (McClatchy et al.,

2007; Wu et al., 2004). The  $^{15}\text{N}$  “heavy” brains served as internal standards for proteome-wide quantitation of transgenic AD mouse models (Tg-AD) relative to non-transgenic (Non-Tg) littermates that were unlabeled and “light”. The mixed “light” and “heavy” proteins were digested to peptides, separated using multi-dimensional chromatography, and analyzed by tandem mass spectrometry (MS) (Figure 1A). We compared the FC, HIP, and CB between age-matched Tg-AD or Non-Tg littermates using the same whole brain “heavy” reference proteome and we performed our analysis at presumably a- (3 mo) and symptomatic (12 mo) time points to identify age dependent changes in protein abundance. In total, our MS analysis provided relative quantitation for > 3,800 different proteins in each dataset of three biological replicates per group (Figure 1B).

By analyzing regional brain proteomes of two Tg-AD models with different degrees of pathology, we identified unknown AD pathological mechanisms and confirmed several previously reported. To identify proximal  $\text{A}\beta$  targets, we first analyzed mice over expressing  $\text{APP}_{\text{swe/ind}}$  (hAPP) at moderate levels. The hAPP mice have hampered synaptic transmission at two to four mo, but do not develop plaques before 10 mo (Hsia et al., 1999; Mucke et al., 2000). We also analyzed a double Tg-AD mouse model over expressing  $\text{APP}_{\text{swe/PSEN1(A246E)}}$  (hAPP/PS1) (Borchelt et al., 1996). hAPP/PS1 mice have plaques, dystrophic neurites, and behavioral deficits by 9 mo (Wang et al., 2003). To document the pathological state of our hAPP mice at 3 mo we performed  $\text{A}\beta_{1-42}$  ELISA and found a significant, ~50%, increase in the level of  $\text{A}\beta_{1-42}$  in the FC and HIP but not CB (Figure S1A). At 12 mo  $\text{A}\beta_{1-42}$  levels in the FC and HIP were more significantly increased compared to the Non-Tg littermates, while the CB did not possess significantly increased  $\text{A}\beta_{1-42}$  levels. The absence of a significant increase in  $\text{A}\beta_{1-42}$  levels in the CB of the hAPP provided a negative control for proteins sensitive to APP fragments outside  $\text{A}\beta$ .  $\text{A}\beta_{1-42}$  ELISA showed a large and significant increase in the  $\text{A}\beta_{1-42}$  level in all three brain regions at 3 and 12 mo in hAPP/PS1 (Figure S1B). We analyzed the amyloid plaque load in HIP by Congo red and Thioflavin S staining in both models and confirmed 3 mo hAPP mice lack plaques, the number of plaques increased with age in both models, and the plaque load even at 12 mo is moderate for both models (Figure S1C–P). Based on these results and previous behavioral analyses we consider the HIP and FC datasets from 3 mo hAPP mice to best represent early stages of pathology.

Our analysis revealed a global imbalance in protein levels, about 5% of the quantified proteins had a change of > 50% at 3 mo and an even larger percentage altered > 50% at 12 mo in 11 out of 12 of our brain region matched comparisons (Figures 1C–D). These measurements were made from, on average 30 different peptides per protein and > 28% of proteins were quantified in 50% of our samples (Figure S1Q–R). The total number of significantly altered proteins increased from 3 to 12 mo in all brain regions and models. Overall, the HIP possessed the greatest number of significantly (FDR-adjusted (adj)  $p$  value < 0.05) altered proteins (Figure 1E). We compared the levels of all the significantly altered proteins in the same Tg-AD model and brain region between the 3 and 12 mo datasets. For FC of hAPP but not hAPP/PS1 mice, we found that the difference between the average protein fold change at 3 to 12 mo was significantly increased ( $-0.0621 + 0.389$  versus  $0.214 + 0.320$ ; mean + STD;  $p$  value =  $1.34 \text{ E}-8$ ), consistent with an accumulation of protein in the context of more severe pathology (Figure S1S–T). For HIP, we found a significant increase

between 3 and 12 mo in hAPP/PS1 ( $-0.266 \pm 0.619$  versus  $-0.155 \pm 0.604$  mean  $\pm$  STD;  $p$  value = 0.0240). For CB we found a significant increase in the hAPP/PS1 model, ( $-0.00866 \pm 0.554$  versus  $0.209 \pm 0.489$ ; mean  $\pm$  STD;  $p$  value =  $3.34 \times 10^{-4}$ ). We determined the number of significantly altered proteins in each brain region and identified those altered in multiple brain regions. Each brain region in both models had 100s of significantly altered proteins (Figures 1F–G and Table S1–2). Bioinformatics to determine the cellular compartments that the altered proteins localize showed they reside in wide range of compartments (Figure S1U). We found 42 proteins significantly altered in FC of both models (Figure 1H and Table S3), and similarly, for HIP we found 184 (Figures 1I and Table S3). To determine how many of our altered proteins have yet to be reported, we performed a literature search, and found 17 FC and 92 HIP unreported in AD or AD models (Table S3). We next used hierarchical clustering to compare protein profiles and found region specific patterns (Figures S2A–C). Principal component analysis showed the 3 and 12 mo FC proteomes of both models are most similar to each other, CB both time points for hAPP or hAPP/PS1 were most similar, and HIP datasets were least similar between the models and time points (Figures S2D–F).

### Connecting proteins to genes that predispose for AD

We tested the connection between genes linked to LOAD and those proteins we found with altered abundance (Bertram et al., 2007). Our screen identified twenty-three significantly altered proteins genetically linked to LOAD (Figure 2A). APOE and GOT1 were altered in six datasets, GSTM1 in five, and TF in four. We extracted the model, time point, and brain region where these alterations occurred. Interestingly, we found that APOE was significantly altered only in those datasets with significantly increased A $\beta$  levels. Other proteins such BIN1 and CLU, were only altered in hAPP/PS1 mice (Figure 2B). These findings show that a large number of genes genetically linked to AD also have altered protein products in mouse model brains of AD-like pathology.

We generated rank ordered summary “confidence plots” for the significant proteins (FDR-adj  $p$  value  $< 0.05$ ). In hAPP mice,  $\alpha$ -spectrin 2, which is known to be altered in AD, was the only protein identified as significantly altered in all four data sets with elevated A $\beta$  (Figure 3A). We found 27 significantly altered proteins in three out of four hAPP datasets, 125 in two, and 691 in one. Similarly, only microtubule associated protein 2 (Map2) was significantly altered in all six hAPP/PS1 data sets, three proteins in five, 20 in four, 36 in three, 214 in two, and 469 in one (Figure S2G). Next, we compared the top 30 proteins (~4%) between the models and found nine proteins met this criterion in both models. These mostly-confidently altered proteins included ApoE, Glial fibrillary acidic protein (GFAP), Solute Carrier Family 12 Member 5 (Slc12a5), neurofilament medium peptide (Nefm), clathrin heavy chain 1, Ankyrin 2 (Ank2), spectrin  $\beta$ -III (Spnb3), aconitase 2 (Aco2), and dynein cytoplasmic 1 heavy chain 1 (Dync1h1). These results are consistent with previous findings that the structural proteins Nefm and Spnb3 are altered in AD brains (Ciavardelli et al., 2010; Masliah et al., 1990). Slc12a5, a major chloride extruder, shifts the effect of GABA, is linked to epilepsy, and may be relevant in AD (Lagostena et al., 2010; Palop and Mucke, 2009). Ank2, GFAP, Aco2, Cnp, and Dync1h1 have also been implicated in AD (Burbaeva et al., 2005; Kamphuis et al., 2014; Lazarov et al., 2002; Mangialasche et al., 2015; Silva et al., 2013; Soler-Lopez et al., 2011; Wang et al., 2003a).

Our finding that ApoE was among the most confidently altered proteins moved us to examine its level relative to APP and A $\beta$ . In 3 mo hAPP mice, APP was increased ~2 fold in all brain regions and more so at 12 mo (Figure 3B). For hAPP/PS1, APP levels were slightly higher at both time points and in all brain regions. When we examined ApoE, which in mice is required for A $\beta$  pathologies (Bales et al., 1997), we found its levels significantly increased in FC and HIP at 12 mo in hAPP. In hAPP/PS1, ApoE had significantly increased levels at both time points in FC and at 12 mo HIP and CB (Figure 3C). Interestingly, ApoE levels were higher exclusively in those brain regions with elevated A $\beta$  levels, since it was unchanged at both time points in CB of hAPP (Figure S1A and Figure 3B). Altogether, ApoE was increased in five out of the five possible 12 mo datasets with the highest A $\beta$  levels.

To specifically explore A $\beta$ 42 dependent proteome remodeling, we performed label-free quantitative proteomic analysis of FC, HIP, and CB extracts at 3 and 12 mo of BRI-A $\beta$ 42 mice (Figure S3A) (McGowan et al., 2005; Park et al., 2008). BRI-A $\beta$ 42 was only detected in mice carrying the transgene and overall levels increased over time (Figure S3B). Our BRI-A $\beta$ 42 proteomics quantified >4,000 proteins in each dataset and identified 100's of proteins with significantly altered levels (Figure S3C and Table S4). As expected, we found no significant changes in APP levels but did detect a significant increase in ApoE. We confirmed increased ApoE levels in BRI-A $\beta$ 42 FC and HIP by WB (Figure 3D). Altogether, these results strongly suggest that A $\beta$ 42 accumulation leads to a concomitant increase in ApoE *in vivo* which likely has unexpected and important pathological consequences.

To test the confidence of our MS based protein measurements we performed semi-quantitative western blot (WB) analyses of six proteins. Indeed, these results unanimously confirmed proteins with both increased and decreased levels (Figure S3D–H). Moreover, BRI-A $\beta$ 42 proteomic analysis also provided additional independent confirmation of many proteins found altered in hAPP and hAPP/PS1. Altogether, these results show that our proteomic analysis possesses sufficient analytical power to reveal confident changes in protein levels relevant to AD pathology.

### Protein co-expression network analysis of Tg-AD brains

We hypothesized consensus weighted gene/protein co-expression analysis (WGCNA) that identifies correlated patterns of protein levels in individual brain regions across the two time points and models, would allow us to delineate groups of co-regulated proteins (Langfelder and Horvath, 2007). We were able to generate consensus networks across genotypes and time points for both the FC and HIP datasets but not CB. We identified 25 FC and 10 HIP MEs across 30 FC and 34 HIP datasets (Figures 4A, S4A, and Table S5). WGCNA reduced 1000s of proteins measurements into 35 MEs that represent core proteome remodeling programs responding to A $\beta$  (Zhang et al., 2013). Direct comparison of the two topological overlap matrices from the Tg-AD or Non-Tg datasets showed A $\beta$  significantly ( $Z$  statistics  $p$  value < 1E-04) remodels several distinct molecular interaction networks (Figures 4B and S4B). To test if the MEs were enriched for distinct functions (Zhang et al., 2013), we subjected each ME to Gene Ontology cell component (GO:CC) enrichment analysis. Indeed, we identified significant gene ontologies assignments for seven of 10 HIP MEs. These

included myelin sheath, regulation of actin polymerization, mitochondrion, synapse, adherens junction, and basal laminae (adj *p* values 6.41E-04 – 1.9E-17) (Table S5), consistent with previous findings (Bartzokis et al., 2007; Terry et al., 1991; Yamaguchi et al., 1992).

By decomposing our MEs into individual time points, we identified shared and unique protein expression patterns between the two models (Zhang and Horvath, 2005), and by using a meta-analysis of correlations of ME eigenproteins (summary expression profile) (Langfelder et al., 2016), we measured the relationship between MEs and each model at two time points. We found that six HIP and two FC MEs had significant (Z statistic and corresponding meta-analysis *p* value < 0.05), differences in abundance for each time point and model (Figures 4C and S4C–D). We found both synapse MEs (ME1 and ME3) in the HIP to have reduced protein levels. However, the myelin sheath (ME2) had only slightly altered expression patterns, except at 12 mo hAPP that showed a significant increase (Figure 4C). Bioinformatics of the FC MEs resulted in significant, but less dramatic eigenproteins (Figures S4D). Interestingly, only myelin sheath and mitochondrion MEs were found in both FC and HIP (Table S5). The strongest associations were in HIP. Noticeably, in hAPP mice, the synapse MEs (1 and 3) were significantly decreased at 3 mo (Figure 4C), which is before A $\beta$  plaques form (Figure S1D and S1K), consistent with previous reports that synaptic deficits are caused by soluble A $\beta$  peptides (Hsia et al., 1999). In contrast, the hAPP/PS1 synapse MEs were only slightly reduced at 3 mo (Figure 4C), but A $\beta$  plaques were already formed (Figure S1E and S1L). The HIP myelin sheath (ME2) also provides a contrasting view of the two models (Figure 4C and S4D). In hAPP there were only slight differences in eigenprotein levels at both ages while in hAPP/PS1 we observed a substantial decrease at 3 mo and more dramatic reduction at 12 mo. Meta-analysis statistics allowed us to rank each ME in two Tg-AD MEs and highlighted related but distinct changes in the proteomes of the HIP and FC.

### Refining MEs to untangle pathology

Interpretation of affected protein pathway perturbations in Tg AD HIP and FC depends on confident ME characterizations. To identify the most confident groups of proteins in our datasets, we first considered the most significant HIP MEs. ME2 and ME4 were the among the most significant and highly populated MEs and GO:CC analysis showed these MEs enriched in myelin sheath proteins (adj *p* value = 2.22E-15; *n* = 213) and mitochondria respectively (adj *p* value = 1.57E-12; *n* = 117). Since these proteins are known to be impaired in AD, it confirms our analysis strategy (Bartzokis, 2004; Hirai et al., 2001). We focused on synapse related HIP ME1 and ME3, and used KEGG database to test if these MEs are significantly enriched in proteins with shared functional assignments (Kanehisa and Goto, 2000). Interestingly, for ME1, we found that long-term potentiation (LTP) to be the predominant assignment (Figure 5A). This result is consistent with previous reports showing that HIP LTP impairments may represent the basis of hampered memory formation in AD patients (Chapman et al., 1999), while ME3 was enriched for the synaptic vesicle cycle. Consistently, when we used the GO: biological processes (GO:BP) database (von Mering et al., 2005), to examine ME1 and ME3, we found these MEs were significantly enriched for synapse plasticity/actin organization and modulation of synaptic transmission/vesicle

transport processes (Figure 5B). These results suggest that synaptic transmission may be altered in HIP by multiple mechanisms in AD-like pathology. To confirm this strategy, we next explored if our MEs relate to previously characterized AD pathology in the FC. We first used KEGG to test if our MEs are enriched for proteins assigned the AD pathway. We found that ME1, ME2, and ME4 significantly enriched in proteins previously associated with AD. To extend these findings, we subjected these three MEs to GO:BP analysis. This analysis showed that these BP terms, namely ME1 - excitatory postsynaptic potential, ME2 – Oxidative-reduction process, and ME4 – hydrogen ion transporter activity, were significantly enriched (Figure S4E) and all of these pathways are linked to AD pathology.

We examined the expression pattern of the individual proteins of the most significant MEs which provided us with insight into the core protein network perturbations. First we inspected the hippocampal LTP module (ME1), which contained proteins such as GluA2 and GluN2A/B (Figure 5C). Overall, these proteins showed progressive and significant reduction in levels in hAPP but were mostly up at 3 and down regulated at 12 mo in hAPP/PS1. For FC and CB in both models, proteins in this module generally had increased levels. We obtained similar results for the ME1 proteins associated with actin organization, a key process for synapse remodeling (Figure 5D). In general, the proteins comprising the second HIP synapse related module (ME3) showed a more robust reduction in levels in hAPP compared to hAPP/PS1 (Figure 5E). ME2 and ME4, from the FC data sets, are involved with mitochondria and had mostly reduced expression in HIP, but increased expression in FC and CB (Figures S4F–G). Finally, we performed network analysis based on the continuous measure of membership and connectivity based on WGCNA to determine the top 50 hub proteins in HIP ME1 and ME3 (Langfelder et al., 2016). These hub protein alterations are similar to the eigenproteins of these MEs and largely reflected function of corresponding MEs. Several core hub proteins identified in ME1, including Wasf1 and Nckap1 (Figures 5F–G), play critical roles regulating spine structure and have been implicated in AD (Ceglia et al., 2015; Govek et al., 2005; Kim et al., 2006; Yamamoto et al., 2001).

### AMPArs as core synaptic complexes altered in AD

WGCNA identified the core AMPAR subunit GluA2 as a top hub gene of ME1 in HIP. Additional GluA2 regulatory proteins such as Nsf and Camk2b (Braithwaite et al., 2002; Kristensen et al., 2011; Shanks et al., 2012), were also identified as hub genes (Figure 5F). This finding suggests GluA2 containing protein networks may represent key potential targets for AD therapeutic intervention. However, GluA2 plays essential synaptic functions in learning and memory and thus the possibility of directly manipulating it is problematic. Thus, we focused on finding key components of AMPAR complexes contributing in AD. To identify the earliest alterations we examined the levels of AMPAR subunits in HIP and FC datasets from hAPP at 3 and 12 mo. At 3 mo, the only significantly altered AMPAR proteins were TARP $\gamma$ -2 in HIP ( $1.253 \pm 0.162$  versus  $0.847 \pm 0.231$  mean  $\pm$  STD,  $p$  value =  $4.29 \times 10^{-4}$ ) and TARP $\gamma$ -3 in FC ( $1.165 \pm 0.0354$  versus  $2.380 \pm 0.139$  mean  $\pm$  STD,  $p$  value =  $7.21 \times 10^{-4}$ ) (Figure S5A). This result contrasts, results at 12 mo that showed both GluA1 and 2 significantly down regulated in HIP (Figure S5B). Based on the key effects of TARPs on AMPAR trafficking and synaptic responses, we tested if TARP $\gamma$ -2/3 levels are altered in the



context of intact AMPARs in AD model brains. Indeed, AMPARs immunopurified from hAPP brains with anti-GluA2 antibodies analyzed by WB and MS showed decreased levels of TARP $\gamma$ -2/3, strongly suggesting AMPARs in AD model brains have altered interactions with TARP proteins (Figures 6A and S5C). To test if AMPAR complexes are remodeled in AD human cortex, we immunopurified AMPARs and analyzed the precipitates with MS and confirmed reduced TARP $\gamma$ -2 levels in the context of human AD pathology (Figure 6B).

Impaired excitatory synaptic transmission and AMPAR dysfunction have been implicated in AD HIP, however the details are unclear (Hsia et al., 1999; Hsieh et al., 2006). Consistently, we observed a significant reduction in the strength of evoked field potentials in hAPP HIP slices relative to control littermates (Figures S5D–F). We hypothesized that reduction of TARP $\gamma$ -2 in AD could lead to destabilization of synaptic AMPARs since TARP $\gamma$ -2 anchors AMPAR subunits to PSD-95, and this interaction can help deliver AMPARs to synapses (Schnell et al., 2002). We wondered if TARP $\gamma$ -2 might affect basal synaptic transmission in hAPP mice. To test this, we injected lentiviruses (LVs) that overexpress TARP $\gamma$ -2 into CA1 in Tg-AD and Non-Tg littermates (Figures 6C–D and S5G). Consistent with previous findings (Schnell et al., 2002), we found overexpression of TARP $\gamma$ -2 had no effect on basal AMPAR EPSCs in Non-Tg animals, but we found that TARP $\gamma$ -2 overexpression resulted in a significant increase in AMPAR EPSCs in hAPP mice (Figures 6E–F). Additionally, we observed no effects on NMDAR EPSCs (Figures 6G–H). These results support TARP $\gamma$ -2 as a potential target for treating in AD.

## DISCUSSION

Proteomic characterization of AD model brains shows elevated A $\beta$  levels cause proteome remodeling in multiple cell types and pathways. Most ApoE is produced by glia and is likely involved in multiple mechanisms important to AD (Holtzman et al., 2012). Our proteomics showed that ApoE levels are increased in brain regions with high levels of A $\beta$  but not in the CB of hAPP, suggesting that it is primarily involved in A $\beta$  clearance rather than production. This result may contrast previous results showing that the APP intracellular domain can drive ApoE gene transcription and may highlight differences between AD models and the complexity of ApoE biology (Liu et al., 2007). Recently, an allele specific role of ApoE in regulating complement C1q protein has an important role in regulating the rate of synaptic pruning (Chung et al., 2016). It is possible that the synaptic protein defects we identified could be due to altered ApoE levels at tripartite synapses.

Many previous reports have described impaired synaptic transmission and an overall reduction in the number of synapses in the HIP of AD models and patients before appearance of plaques (Shankar et al., 2008; Terry et al., 1991). We confirmed distinct components of excitatory synapses are altered prior to plaque formation. WGCNA revealed two synaptic MEs with reduced protein levels at early and late time points. ME1 is enriched in proteins involved with LTP and actin cyto-skeleton, both of which have been implicated as playing key roles in AD (Sheng et al., 2012). ME3 is enriched in proteins involved with pre-synaptic functions. Among the top 50 hub proteins in ME1 were AMPAR and NMDAR subunits. Examination of AMPAR components showed a pioneering reduction in protein abundance in the auxiliary subunits TARP $\gamma$ -2/3. It was not until the later time point that

GluA1 and 2 had significantly reduced levels in the HIP of hAPP mice. AMPAR density was shown reduced at perforated synapses and synapses onto parvalbumin (PV)-positive neurons in the CA1 region of TARP $\gamma$ -2 KO mice interestingly we found significantly reduced levels of PV in the HIP of both models (Yamasaki et al., 2016). Furthermore, pore forming GluA3 and -4 subunits were not significantly altered. In other AD models they have been showed to be important A $\beta$  substrates (Reinders et al., 2016). AMPAR downscaling and removal have been reported in multiple AD models, but the precise mechanism(s) have remained vague (Chang et al., 2006; Hsia et al., 1999; Hsieh et al., 2006). These findings motivated us to test whether restoring TARP $\gamma$ -2 levels could rescue AMPA defects in hAPP mice. Indeed, over expression of TARP $\gamma$ -2 caused a large increase in AMPAR- but not NMDAR-mediated currents. Our results suggest that GluA1, 2 containing AMPARs that are known to be enriched with TARP $\gamma$ -2/3 may represent key A $\beta$  targets due to their restricted expression patterns at the effected synapses (Schwenk et al., 2012). Finally, while both NMDARs and AMPARs play key roles in the establishment and maintenance of LTP, our results highlight a potential mechanism by which reduced expression of TARP $\gamma$ -2 could result in impaired delivery of AMPARs to spines and thus compromise LTP in AD patients. TARP subtypes have been shown to differentially influence AMPAR gating (Milstein et al., 2007), and TARP $\gamma$ -2 may specifically represent a therapeutic target to restore cognitive function in AD patients.

We determined protein substrates and mechanisms of A $\beta$  toxicity to dissect the amyloid cascade hypothesis by using three complementary mouse models of AD-like pathology (Table S6). To maximize the accessibility our results, we have generated an online interactive AD model protein expression database (Proteomics INTegrator) as a resource for the entire AD research community: <http://sealion.scripps.edu/pint/?project=3d7c1ac078930a798a07c6a397bd21ef>. This resource allows users to query proteins of interest, visualize quantified peptides, and perform enrichment analyses (Figure S6). The importance of determining protein abundances is particularly relevant for the investigation of AD because altered proteostasis and reduced protein degradation capacities are hallmarks (Powers et al., 2009; Rubinsztein, 2006). However, our study is not without limitations since we failed to accurately quantify many proteins expressed at low levels, and our results are limited by protein abundance averaging between the multiple cell types present in the brain.

Our results show that A $\beta$  causes a broad and progressive alteration in the level of many functionally and spatially linked components of the brain proteome. These findings, and those recently published from others (Seyfried et al., 2017), raise the possibility that studying groups of co-expressed proteins might be advantageous over the study of individual proteins due to the highly complex response of various cell types to toxic A $\beta$

## EXPERIMENTAL PROCEDURES

### Mice

Ten C57BL6 female mice were metabolically labelled with <sup>15</sup>N-rich, Spirulina-based diet (Cambridge Isotopes) for 12 weeks starting at P21 (Wu et al., 2004). The <sup>15</sup>N protein enrichment was calculated based on the shape of the peptide isotope envelope, and in the brain was determined to be 90–95% (MacCoss et al., 2005). For the hAPP model

we used the less aggressive J9 line, except for the AMPAr immunoprecipitation – mass spectrometry/western blots and whole cell electrophysiology in which out of necessity we used J20 mice (RRID:IMSR\_JAX:004662) (Mucke et al., 2000). All hAPP mice were on C3H/B6 background. The hAPP/PS1 mice used were (B6C3-Tg(APP<sup>swe</sup>,PSEN1<sup>ΔE9</sup>)85Dbo/Mmjax–129x1/SvJ) (RRID:MMRRC\_037564-JAX) (Borchelt et al., 1996; Cohen et al., 2009). The BRI-Aβ42 mice were C57BL/6J (RRID:IMSR\_JAX:007182) (McGowan et al., 2005). Institutional Animal Care and Use Committees of UCSD, Salk Institute, and the Scripps Research Institute, approved all animal procedures, and studies. Most of the Tg-AD mice used in this study were female, 3 mo = 3.0 – 3.6 mo and 12 mo = 12.2 – 15.0 mo. Brains were split into two hemispheres, one for proteomic analysis and ELISA, and the other for histology. Mice were from the Maslah, Koo, and Dillin lab colonies or JAX. See Supplemental Experimental Procedures for details.

## MS Analysis

Brain region homogenate corresponding to 100 μg, based on BCA assay was digested to peptides with trypsin overnight and was processed for MudPIT as previously described (Savas et al., 2015). The peptide loaded column was placed in line with an Agilent 1200 HPLC and analyzed with an 11 step method. MS analyses were completed with Thermo Orbitrap Velos or Orbitrap Fusion Tribrid mass spectrometers (see Table S6). Protein identification and quantification and analysis were done with ProLuCID, DTASelect2, Census, and QuantCompare within the Integrated Proteomics Pipeline (IP2) environment. Tandem mass spectra were matched to sequences using the ProLuCID algorithm with 50 ppm peptide mass tolerance for precursor ions and 400 ppm for fragment ions. Each dataset had a ~1% FDR rate at the protein level based on the target-decoy strategy. Since we analyzed brain tissue which has many cell types and mRNA alternative splicing we chose large databases to maximize our analysis potential. All the peptide identifications have been mapped to genes and the details can be found in our online database where all proteins were also mapped to UniprotKB accessions.

For MS1 based quantification “Census” first calculates the elemental compositions and corresponding isotopic distributions for both the unlabeled and labeled peptides. All isotopes with greater than 5 % of the calculated isotope cluster base peak abundance were used. MS1 files were used to generate chromatograms from the m/z range surrounding both the unlabeled and labeled precursor peptides. The core of Census is a linear least-squares correlation that is used to calculate the ratio (i.e., slope of the line) and closeness of fit [i.e., correlation coefficient ( $r$ )] between the data points of the unlabeled and labeled ion chromatograms. In this study, only peptide ratios with the coefficient correlation values ( $r^2$ ) greater than 0.5 were used for further analysis. The Grubbs test ( $p$  value < 0.01) was used to remove outlier peptides. Final protein ratios were generated with QuantCompare, which uses Log two-fold change on the biological replicates. The statistical significance labeled of the differential expression of all proteins was assessed using a two-tailed one-sample t-test on their corresponding peptide quantification ratios between both conditions. The obtained  $p$  values were FDR-adjusted for multiple hypothesis testing using the Benjamini–Hochberg correction (Benjamini and Hochberg, 1995). Proteins with FDR-adjusted  $p$  values < 0.05 and for which quantification measurements were obtained in at least two biological

replicates in both conditions were considered for further analyses. We also performed label-free quantitative analysis using Census within IP2. See Supplemental Experimental Procedures for details.

### Statistical and Bioinformatic Analyses

We compared our proteomic data to the previously described potential Alzheimer disease susceptibility genes (Bertram et al., 2007). For all heat maps, we used 3 biological replicates for each experimental group, required each protein to be quantified in a minimum of 57 (out of a total of 72) datasets and had to be quantified in each brain region. We used the Genes to Cognition protein database to identify synaptic proteins, MitoCarta2.0 for mitochondrial proteins, and InnateDB to identify the immunological proteins. Box plots (Figure 3) define 25th and 75th percentile and statistical analysis were performed with student's t-test with 2 tails and un-equal variance. For quantification of APP levels, only peptide sequences with 100 % homology between human and mouse were used. Bar graphs (Figure S1) show mean  $\pm$  SD with student's t test with 1 tail and un-equal variance (A & B) with 2 tails and equal variance (I & P). Bar graphs for human AMPAR purifications with MS analysis, statistics were performed with Student's t test with 1 tail and 2 sample equal variance (Figure 6B). Western blot quantification Bar graphs (Figure S3) show mean  $\pm$  SEM and student's t test with 1 tail and two-sample equal variance. We used the program 'Venn Diagram Plotter' to construct two and three circle Venn diagrams (aka Euler diagrams). Detailed description of each analysis, the tools used, and the specific criteria is in the Supplemental Experimental Procedures.

### WGCNA

The consensus weighted correlation networks were constructed across both Non-Tg and Tg-AD groups for each brain regions with R (Langfelder and Horvath, 2008; Zhang and Horvath, 2005). Prior to analysis, protein IDs with 14 or more missing values were removed, resulting in three pairs of datasets: 1912 protein IDs for HIP, 2146 protein IDs for FC and 2169 protein IDs for CB. The specific datasets for each analysis is as follows (dataset numbers correspond to the numbering indicated in Table S6): FC = 137-9, 124-7, 89-92, 77-9, 112-6, 101-3, 65-7, 53-7. HIP = 140-4, 128-31, 93-5, 80-4, 117-9, 104-8, 68-72, 58-61. The soft power threshold was set to 12 for HIP groups, 6 for FC groups to arrive at the network adjacency. See Supplemental Experimental Procedures for details.

### Affinity purification of AMPA receptor complexes

We used our previously reported method to purify native AMPARs (Nakagawa et al., 2005). Human brain (cortex) was obtained through the National Disease Research Interchange (NDRI), Researcher: Yates (code YAJ2), TSRI: IRB-11-5719. The antigen of the antibody against GluA2 is conserved in rat and human. See Supplemental Experimental Procedures for details.

### Electrophysiology

Brains were cut into 300  $\mu$ m sagittal sections on a vibratome and placed in ice-cold carbogenated artificial cerebrospinal fluid. Whole-cell recordings were made using 3-5 M $\Omega$

pipettes filled with an internal solution that contained 123 mM Cs-gluconate, 8 mM NaCl, 1 mM CaCl<sub>2</sub>, 10 mM EGTA, 10 mM HEPES, and 10 mM glucose, pH 7.3 with CsOH, 280–290 mOsm. Responses were evoked by stimulating axons in the stratum radiatum with a platinum 2-contact cluster electrode 100–200 microns lateral to the recording site. See Supplemental Experimental Procedures for details.

## Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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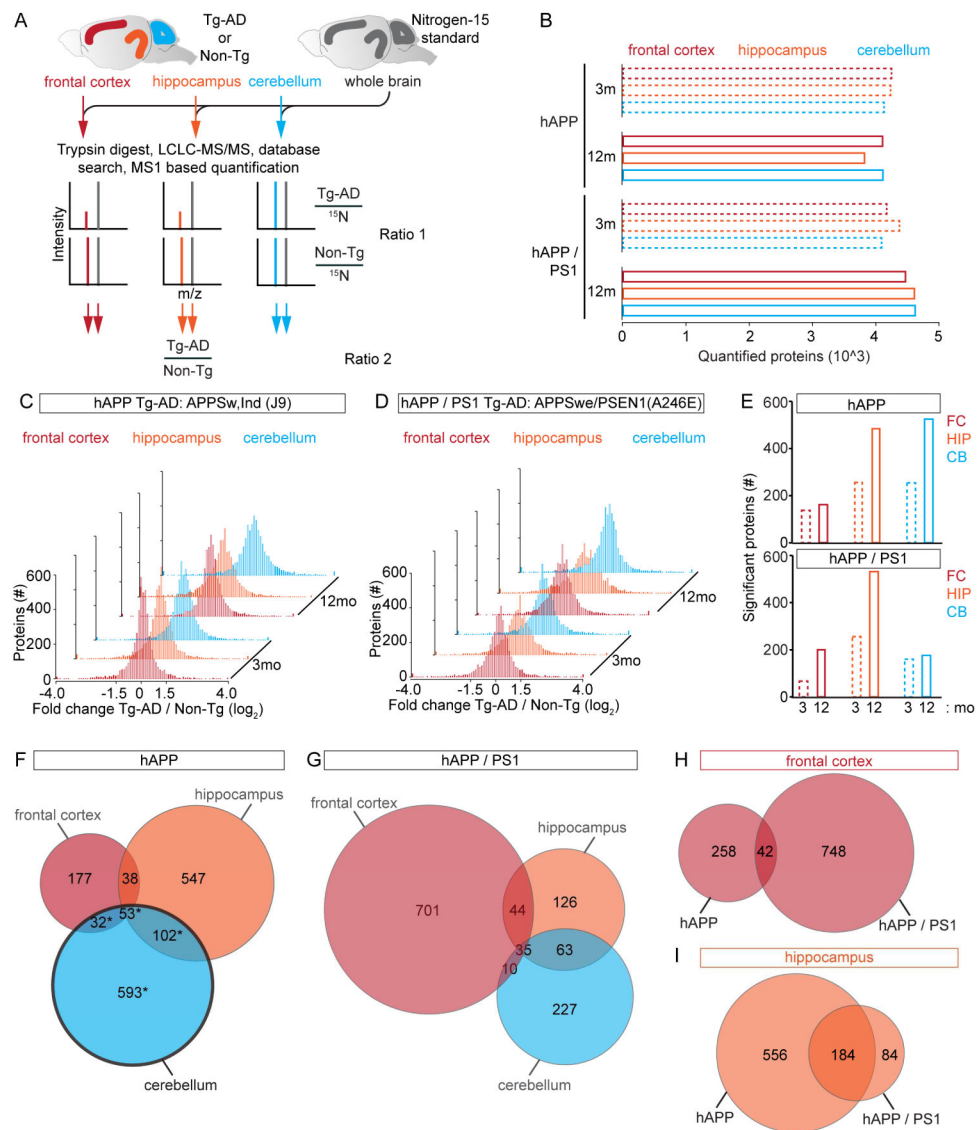
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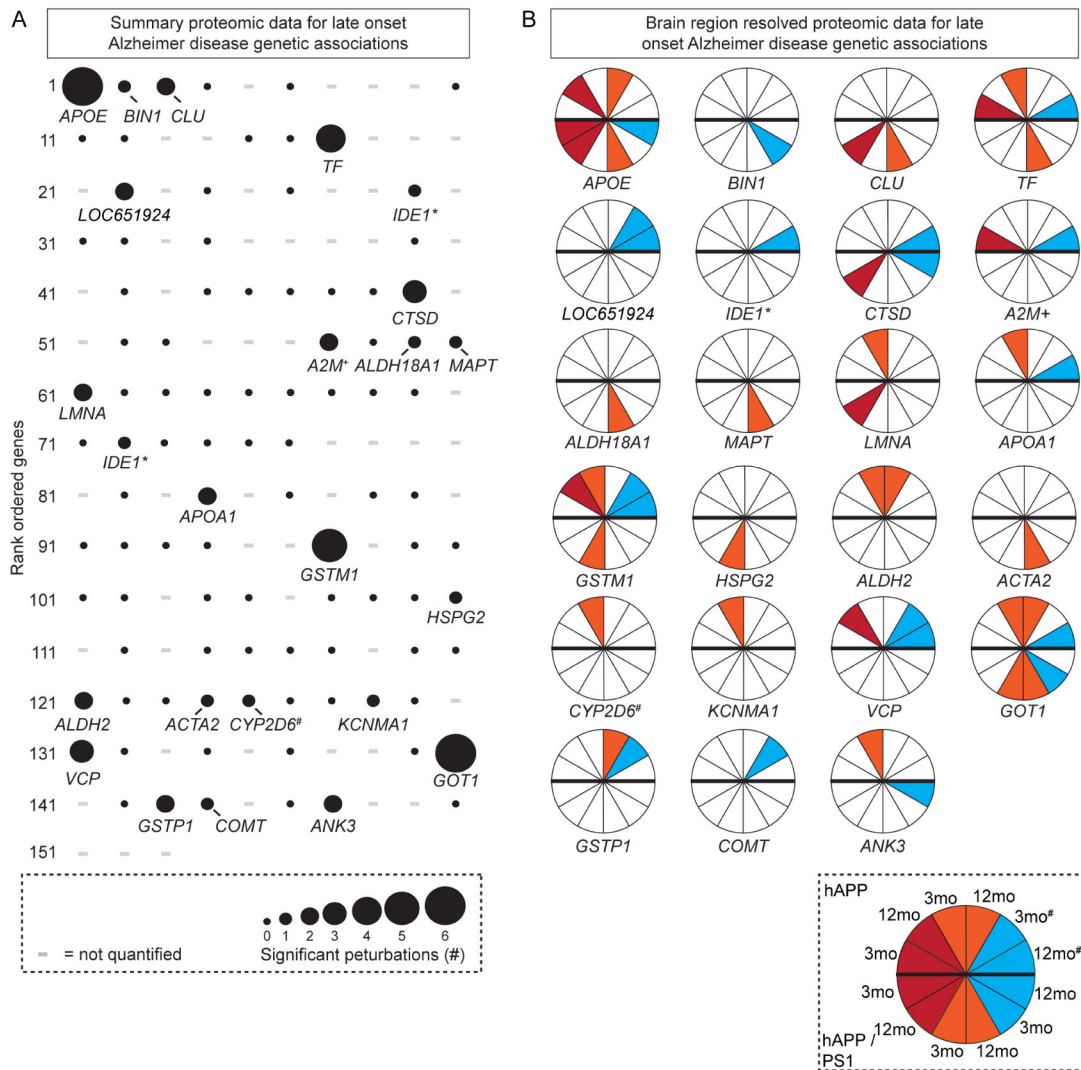
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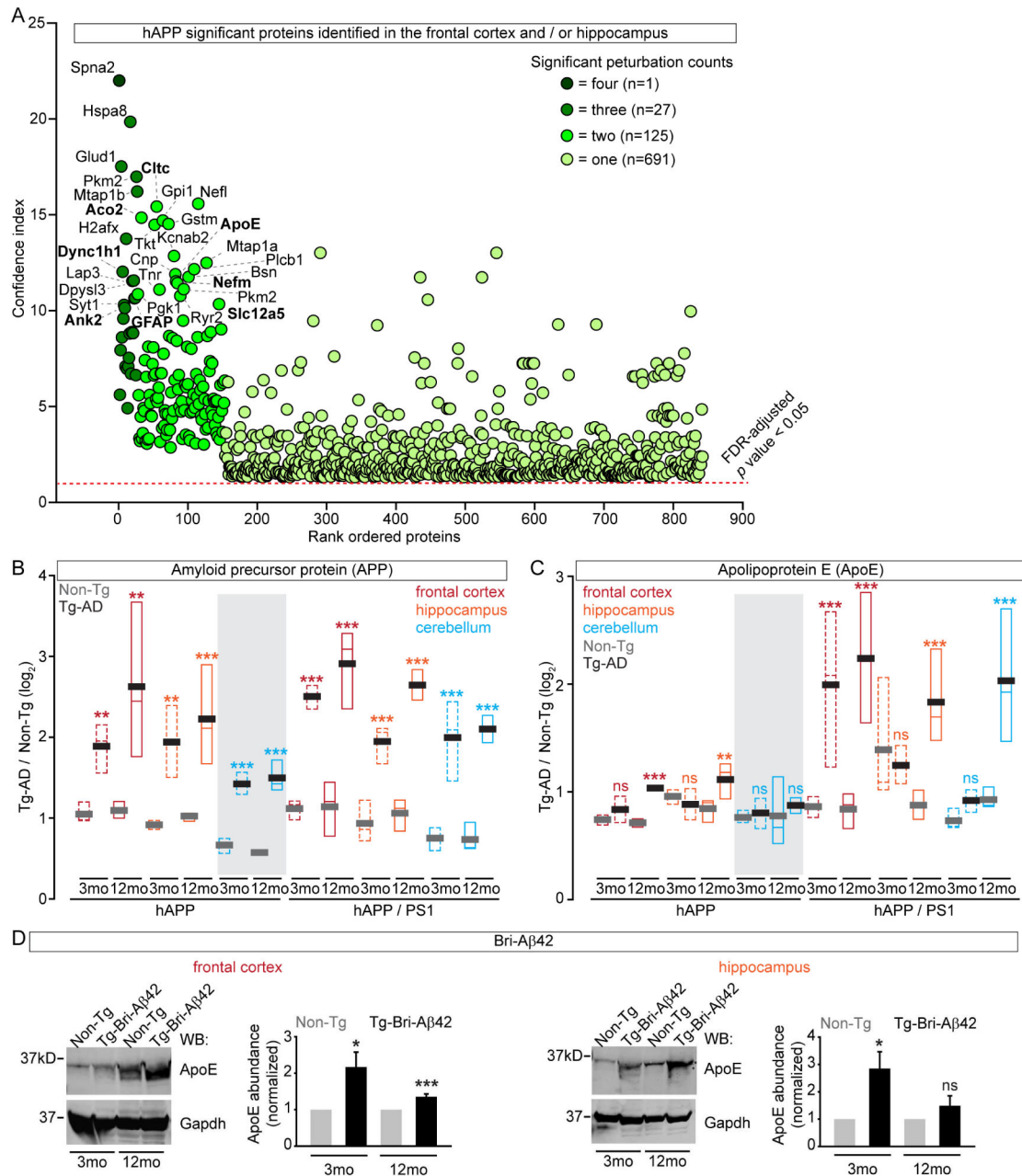


**Figure 1.** Quantitative MS Analysis of Tg-AD Brains **(A)** Experimental workflow,  $^{14}\text{N}$  ratio  $^{15}\text{N}$  is calculated and normalized to controls. **(B)** Quantified protein counts for 12 datasets. **(C)** Brain region proteome remodeling of hAPP at 3 and 12 mo. **(D)** Brain region proteome remodeling of hAPP/PS1 at 3 and 12 mo. **(E)** Bar graphs of age dependent increase of the number of significantly altered proteins (FDR-adj  $p$  value < 0.05). **(F–G)** Venn diagrams of significantly altered proteins. hAPP black outline = CB lacking increased  $\text{A}\beta$  levels. **(H)** FC summary for both models, **(I)** HIP summary for both models. For (B and C) Solid bars are proteins with a > 50% change in ratio 2. See also Figure S1, S6 and Table S1–3, 6.



**Figure 2.**

Altered Protein Abundance of Genes Linked to LOAD (A) Significant (FDR-adj  $p$  value < 0.05) protein changes mapped to 153 genes previously linked to LOAD. - = not quantified. (B) Brain region(s) and age(s) with significant changes. Red = FC, orange = HIP, and teal = CB. # = hAPP CB that lacks A $\beta$  accumulation. \* = IDE1\* is shown twice because of sequence variations, + = A2M+ was not identified - related alpha-2-macroglobulin protein Pzp is reported, # = CYP2D6 is not in the mouse genome, we measured related Cyp2d22.



**Figure 3.**

Most Significant Altered Proteins from Brain Regions with Elevated A $\beta$  (A) Summary confidence index (see methods) plot for all significantly (FDR-adj  $p$  value < 0.05) altered proteins in the four datasets with significantly increased A $\beta$  levels in hAPP. Proteins with small  $p$  values > 0 but < 3.42 E-09 graphed as 1E-13 (see Tables S1); proteins in the top 30 of both models are in bold. (B) APP levels in Tg-AD datasets revealed a significant increase compared to Non-Tg controls (n = 8 – 32 peptides). (C) Apo E levels in hAPP are significantly increased at 12 mo in FC and HIP but not CB. In hAPP/PS1 ApoE levels are significantly increased at 3 mo in the FC, and at 12 mo in all three brain regions (n = 17 – 39

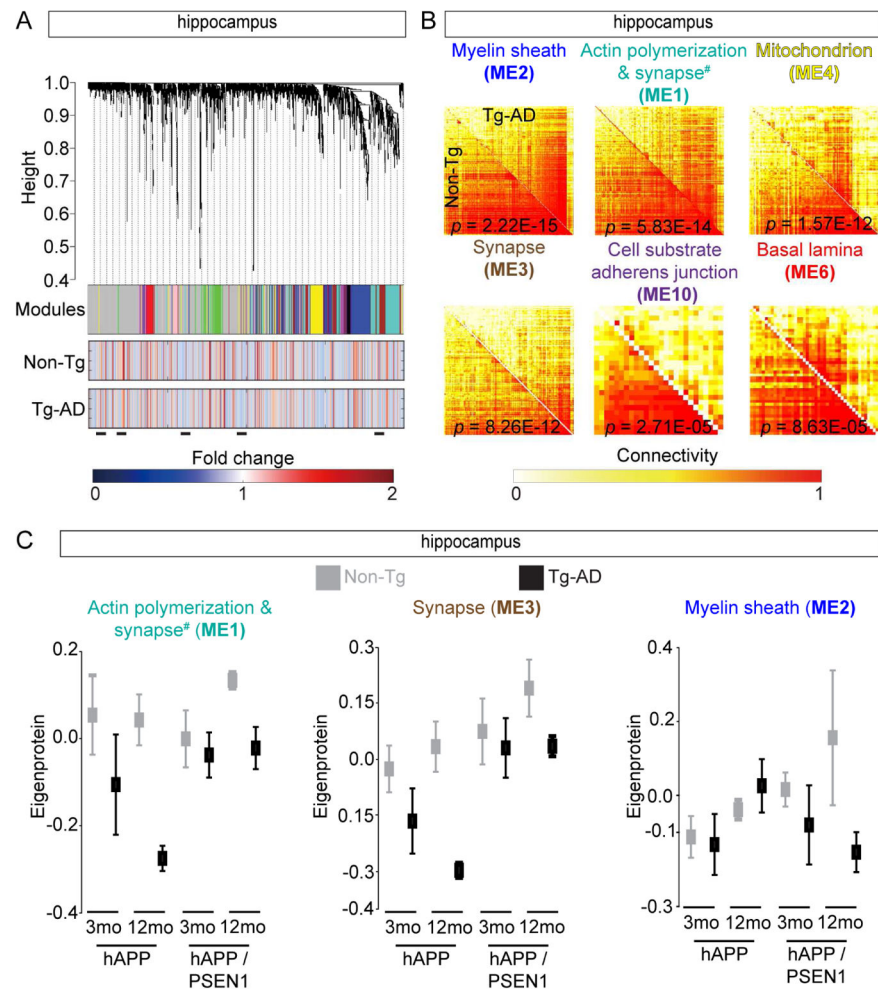
peptides). **(D)** Apo E is increased in BRI-A $\beta$  FC and HIP by WB. Median = colored, mean = grey or black, grey shading = datasets without increased A $\beta$  levels. \* $p$  value < 0.05, \*\* < 0.01, \*\*\* < 0.001 by Student's t test for (B–D). See also Figure S2–3 and Table S4.

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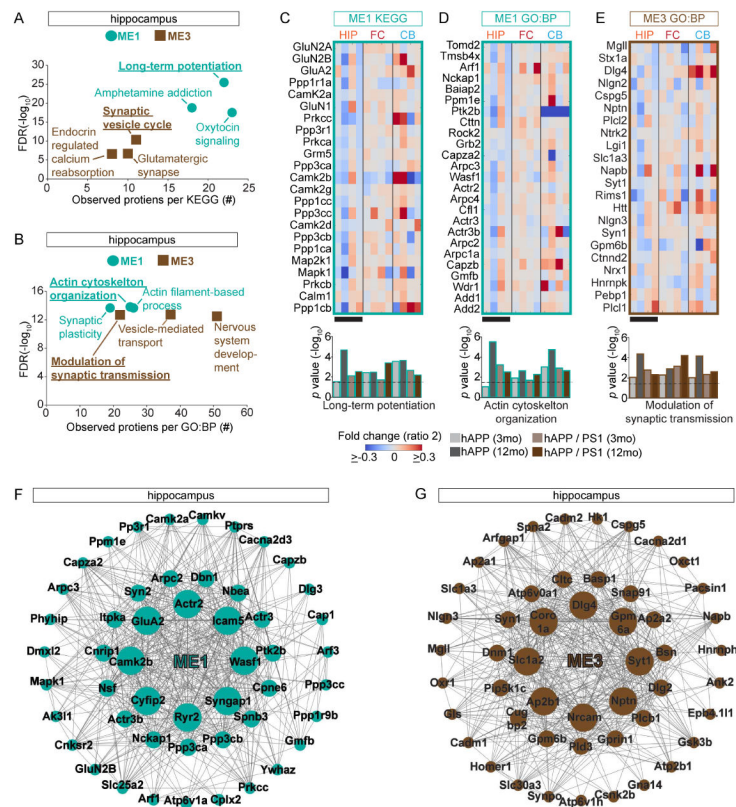
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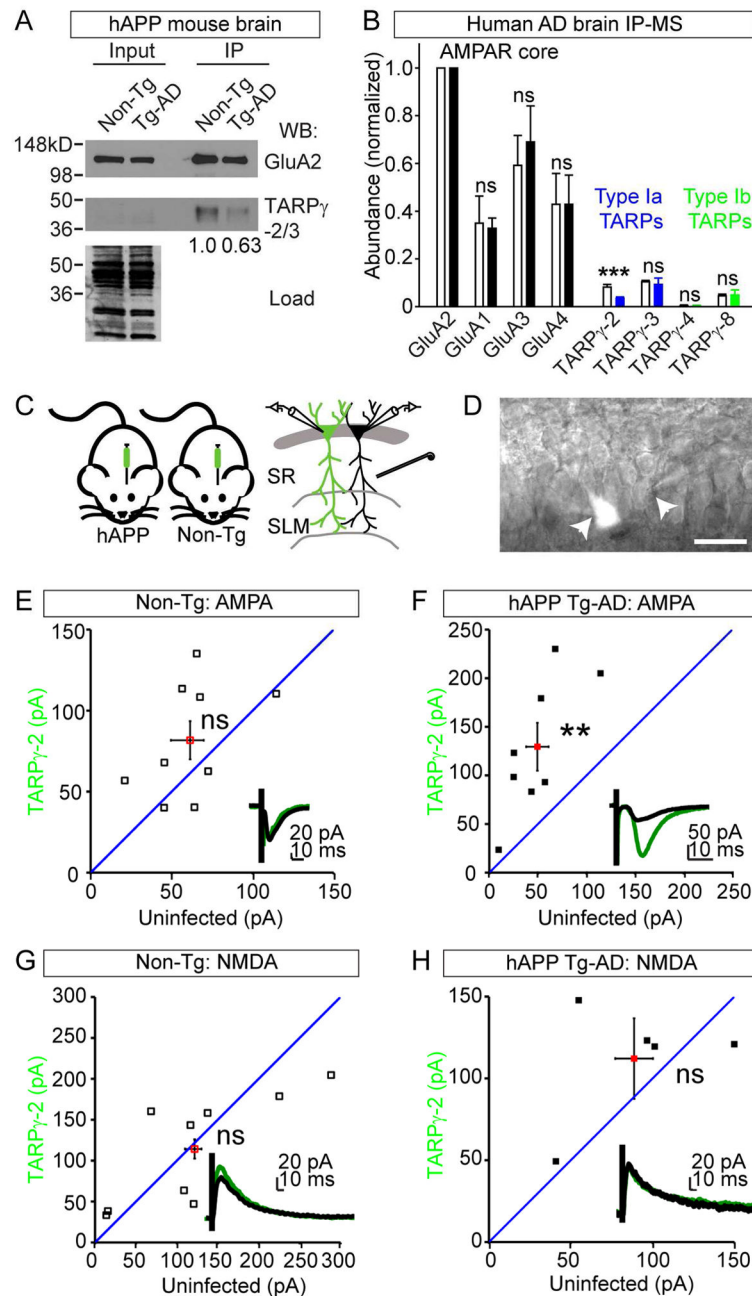
**Figure 4.**

Consensus Protein Coexpression Network Analysis of Tg-AD Brains (A) HIP protein clustering trees from both Tg-AD models and time points (top). Each module is in non-grey (middle). The two rows of heat maps below show the association of individual proteins with  $Q$  for Non-Tg or Tg-AD groups (bottom). Blue and red shading indicates proteins with reduced or increased expression, respectively, with increasing  $Q$ . Black bars show altered protein groups. (B) Individual topological overlap matrices of significantly ( $Z$  statistics  $p$  value  $< 1E-04$ ) differentially connected modules in HIP between Tg-AD and Non-Tg datasets in both models. Module assignment based on most significant GO assignment (# for ME1 see Table S5). Shown are all modules with significant  $p$  values ( $< 1E-4$ ). (C) Summary expression value (eigenprotein) from the indicated HIP datasets, all Non-Tg/Tg-AD comparisons have  $p$  value  $< 0.05$  calculated from  $Z$  statistics. See also Figure S4 and Table S5.



**Figure 5.**

Synaptic Proteins in ME1 and ME3 Have Specific Changes in Protein Levels (A) HIP ME1 and ME3 are enriched for proteins with distinct synaptic functions based on KEGG. Scatter plot showing enrichment FDR (y-axis) versus the number of proteins per KEGG pathway (x-axis). (B) HIP ME1 and ME3 are enriched in proteins with distinct synaptic functions based on GO:BP. (C–E) Protein expression matrix (ratio 2) for the indicated MEs for both Tg-AD models and time points. Black bars show datasets with contrasting protein expression patterns. Below bar graph based on hypergeometric enrichment  $p$  values, dotted line indicates an adj  $p$  value  $< 0.05$ . (F–G) Network of top 50 hub proteins in ME 1 and ME3 from the HIP (8 most connected = large, next 16 = medium, remaining 26 = small). (A and B) Most significant KEGG and GO:BP for ME1 and ME3 are underlined. See also Table S5.



**Figure 6.** AMPAR Complexes Are Hub Proteins Altered in AD (A) WBs of AMPAR complexes from hAPP or Non-Tg brains. Less TARP- $\gamma$ 2/3 was recovered in the hAPP brain compared to Non-Tg after normalizing to the recovery of GluA2: 1.0 versus 0.63. (B) Semi-quantitative MS analysis of AMPAR complexes from AD or healthy control brains show AMPARs in AD cortex have reduced levels of TARP- $\gamma$ 2 associated normalized to GluA2: healthy cortex =  $0.082 \pm 0.011$  versus AD cortex =  $0.035 \pm 0.0058$ . Bar shows mean  $\pm$  standard deviation, AD brains (n = 4 patients) and controls (n = 2). White bars = controls and solid bars = AD. (C) Strategy to test if injected LVs expressing Flag-TARP- $\gamma$ 2-IRES-GFP can rescue AMPA

defect in hAPP mice. **(D)** Image of simultaneously recorded un- and infected CA1 cells in a hAPP HIP brain slices. **(E)** AMPA currents are not significantly different in Non-Tg mice (TARP- $\gamma$ 2:  $70.248 \pm 11.479$ ; Uninf:  $59.896 \pm 11.020$ ,  $n = 7$ ,  $p$  value = 0.379). **(F)** AMPA currents are significantly increased in TARP- $\gamma$ 2 expressing cells in hAPP mice (TARP- $\gamma$ 2:  $134.67 \pm 27.812$  pA; Uninf:  $48.331$  pA,  $n = 7$ ,  $p$  value = 0.00385). **(G)** NMDA currents are not significantly different in Non-Tg mice (TARP- $\gamma$ 2:  $114.2344 \pm 22.5515$ , Uninf:  $121.669 \pm 29.973$ ,  $n = 9$ ,  $p$  value = 0.7089). **(H)** NMDA currents are not significantly different in hAPP mice (TARP- $\gamma$ 2:  $112.148 \pm 16.537$ ; Uninf:  $88.6866 \pm 0.301$ ,  $n = 5$ ,  $p$  value = 0.3016). (E–H) Mean  $\pm$  SEM. \*\* $p$  value < 0.01, \*\*\* < 0.001, by Student's t test. See also Figure S5.