Tau-induced elevation in promoter-proximal RNA polymerase II pausing is linked to decreased expression of long neuronal genes in a *Drosophila* tauopathy model.

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Abstract

Tauopathies, including Alzheimer's disease, are age-related neurodegenerative disorders characterized by abnormal phosphorylation and buildup of microtubule-associated protein tau. Gene expression dysregulation is a key molecular feature of tauopathies, but how aging and disease interact to disrupt crucial transcriptional regulators and pathways remains largely unknown. Here, we examined how pathological tau affects gene expression programs in age-related neurodegenerative disease using a well-established *Drosophila melanogaster* tauopathy model with neuronal expression of the toxic human tau[R406W]. Transcriptomic analysis of tau-expressing fly heads showed a preferential downregulation of long neuronal genes with long introns. Notably, we found that these downregulated genes in the tauopathy model are marked by increased accumulation of initiating RNA polymerase II (RNAP II) near the transcription start site and reduced elongating RNAP II within gene bodies, indicating a problem with the transition from initiation to elongation. By calculating an RNAP II Pause Index (PI) for each gene, we identified a strong link between promoter-proximal RNAP II stalling, gene expression deficits,

28   and gene length in the tauopathy model. Overall, we have uncovered the genomic and

29   transcriptomic features of tau-dependent downregulated genes and identified increased RNAP II

30   promoter-proximal stalling as a significant mechanism of transcription stress in tauopathy.

1. Introduction

Tauopathies are a diverse group of age-related neurodegenerative disorders characterized by the

abnormal accumulation of tau, a microtubule-associated protein essential for microtubule

assembly and stability (Black et al., 1996; Ramírez et al., 1999). The presence of intracellular

neurofibrillary tangles composed of hyperphosphorylated tau is a key pathological feature of

Alzheimer's disease (AD) and frontotemporal dementia with parkinsonism linked to

chromosome 17 (FTDP-17) (Hutton, 2000). Mutations in the microtubule-associated protein tau

(MAPT) gene that affect microtubule binding, phosphorylation, or aggregation are responsible

for most FTDP-17 cases (Goedert and Spillantini, 2000). FTDP-17 primarily affects the frontal

and temporal lobes of the brain, leading to progressive changes in behavior and personality,

communication difficulties, and impairments in memory and motor skills.

Although tau is involved in the development of various neurodegenerative diseases, the cellular

mechanisms that cause neuronal dysfunction and death are still not well understood.

Understanding the pathways that promote tauopathy progression, especially in the context of

aging—which is the main risk factor for the disease—is crucial for deepening our knowledge of

the molecular basis of tau-driven neurodegeneration (Niccoli and Partridge, 2012). Gene

expression is increasingly recognized as a key factor in the development of tauopathies in both

mouse models and human patients (Miller et al., 2008, 2010; Tan et al., 2021). During aging,

existing transcriptional programs undergo widespread changes, driven by factors such as

accumulated DNA damage and epigenetic modifications (López-Otín et al., 2023). These

changes are thought to cause many of the physical signs of aging and age-related diseases.

Previously, we showed that long, highly expressed genes encoding proteins with neuronal

functions exhibit reduced expression in aging neurons (Hall et al., 2017). Later studies have also

supported a connection between gene structure, transcription levels, and cellular function across

different tissues and organisms, in aging and neurodegenerative disease contexts (Barbash and

Sakmar, 2017; Lopes et al., 2021; Jauregui-Lozano et al., 2022; Stoeger et al., 2022; Debès et al.,

2023; Gyenis et al., 2023). However, how aging and disease interact to alter the cellular environment and affect key transcriptional regulators that control gene expression remains largely unknown.

To investigate how tau affects gene expression programs in age-related neurodegenerative disease, we used a well-established *Drosophila melanogaster* tauopathy model with neuronal expression of human tau carrying the R406W mutation. Using steady-state RNA sequencing (RNA-seq), we characterized the age- and tau-dependent transcriptome in tauopathy flies. In tau transgenic flies, we observed a preferential downregulation of long genes characterized by long introns that encode factors essential for neuronal structure and function. Additionally, using the RNA polymerase II (RNAP II)-targeted Cleavage Under Targets and Release Using Nuclease (CUT&RUN) approach, we analyzed tau-dependent transcription dynamics. Downregulation of long genes was strongly linked to increased RNAP II stalling near the transcription start site (TSS) and decreased RNAP II occupancy along the gene body in the tauopathy model, suggesting that impaired promoter-proximal pause release is a likely mechanism behind their reduced expression. Overall, our results reveal the transcriptional dynamics and genomic features of genes dysregulated in tauopathy and their relationship to progressive, age-related neurodegeneration. Importantly, promoter-proximal RNAP II stalling emerged as a key characteristic of long, downregulated neuronal genes, implicating tau-dependent dysregulation of the transition from transcription initiation to productive elongation as a possible molecular driver of transcriptional stress in tauopathy.

2. Materials and Methods

2.1 *Drosophila* Stocks and Maintenance

The following strains used in this study were from Bloomington *Drosophila* Stock Center: elav>Gal4 (BDSC #8765), UAS-lacZ (BDSC #8529), UAS-tau (BDSC #78851). The UAS-TAU^{R406W} strain was a gift from Mel Feany (Wittmann et al., 2001). Flies were maintained under a 12:12-hour light: dark cycle at 25°C on standard fly food. For aging experiments, flies were collected within three days post-eclosion and transferred to fresh food every two days.

2.2 Optic Neutralization

Optic neutralization was performed as previously described (Stegeman et al., 2018). Briefly, 10

male flies per genotype were aged to the designated time point, anesthetized on a CO_2 pad, and

mounted to glass slides using clear nail polish as adhesive. Live rhabdomeres were imaged using

brightfield light microscopy at 20x magnification. For each genotype and time point, at least 70

rhabdomeres were assessed. Retinal degeneration was quantified in a blinded manner and

expressed as the percentage of lost rhabdomeres relative to the total number of rhabdomeres.

2.3 Climbing Assay

Climbing assays were performed using groups of 25 male flies per genotype. Flies were sorted

into vials and allowed to recover for 24 hours following anesthesia. The next day, flies were

gently tapped into a 100 ml glass graduated cylinder and acclimated for 2 minutes. The cylinder

was then tapped firmly and swiftly three times to bring the flies to the bottom. Flies were given

10 seconds to climb, and the number of flies above a 5 cm mark was recorded. Each biological

replicate consisted of three technical trials, with a one-minute rest between tests. For each

genotype, 6-12 biological replicates (a total number of 150-300 flies) were assessed.

2.4 Western Blotting

Western blot analysis was performed using 40 μg of whole-cell protein extract from 25 fly heads

per genotype. Heads were manually homogenized in cell lysis buffer (50 mM Tris-HCl, pH 8.0,

150mM NaCl, and 1% NP-40). Antibodies used for detection are listed in Supplementary Data 2.

2.5 RNA Extraction, Reverse Transcription-quantitative PCR (qPCR), and RNA-seq

Total RNA was extracted from 20-25 fly heads per genotype by mechanical homogenization in

TRIzol reagent (Thermo Fisher, #15596018), followed by phenol/chloroform extraction and

ethanol precipitation using standard protocols. cDNA was synthesized from 300-400 ng of total

RNA using Episcript RNaseH Reverse Transcriptase (Biosearch Technologies, #ERT12925K-

ENZ) with random hexamer primers (IDT). Quantitative PCR (qPCR) was performed using

PR1MA qMAX Green 2X qPCR mix (Midsci, #PR2000-N-5000) on a CFX Connect Real-Time System (BioRad) according to the manufacturer's instructions. Relative gene expression was normalized to the geometric mean of two reference genes in accordance with MIQE guidelines (Bustin et al., 2009). Primer sequences are provided in Supplementary Data 2.

For RNA-seq, extracted RNA was treated with DNase I (Roche, #04716728001) to remove genomic DNA according to the manufacturer's protocol. Purified RNA was quantified using the Qubit RNA HS Assay Kit (ThermoFisher, #Q32852). cDNA libraries were prepared from 250 ng of RNA using the Universal Plus™ Total RNA-Seq Library Preparation Kit, incorporating *Drosophila*-specific anyDeplete technology for rRNA depletion (Tecan, #9156). Libraries were pooled and sequenced using paired-end 150 bp reads in a single lane on an Illumina Novaseq X Plus platform.

2.6 Bioinformatic Analysis of RNA-seq Data

Bioinformatic analysis of RNA sequencing data was performed as previously described (Jauregui-Lozano et al., 2022) with minor adjustments. Raw reads were trimmed using Trimmomatic (Bolger et al., 2014) to remove adapter sequences and low-quality bases. Cleaned reads were then quantified using Salmon (Patro et al., 2017) against a custom reference transcriptome built from *Drosophila melanogaster* (BDGP6.54) cDNA sequences, with the addition of *Homo sapiens* MAPT transcript information extracted from the GRCh38 cDNA reference. The resulting transcript-level counts were summarized to gene level and analyzed in R using the DESeq2 (Love et al., 2014) package for differential expression analysis. Cut-offs for differential gene expression were adjusted to a p-value < 0.05 and |fold change| > 1.5 unless otherwise stated. For comparisons between differential expression groups, statistical analysis was performed using ANOVA followed by a Tukey HSD post-hoc test. Gene Ontology (GO) enrichment analysis of significantly differentially expressed genes was performed using clusterProfiler (Yu et al., 2012).

2.7 CUT&RUN

CUT&RUN was performed using the EpiCypher CUTANA protocol with modifications for

nuclei isolation from fly head tissue. Briefly, 100 fly heads were homogenized in 1 mL of

Nuclear Extraction Buffer (20 mM HEPES, pH 7.9, 10 mM KCl, 0.1% Triton-X, 20% glycerol,

0.5 mM spermidine, 1× protease inhibitor) using a 1 mL Dounce homogenizer. Homogenization

comprised 7 loose pestle strokes, a 5-minute ice incubation, followed by 7 loose and 12 tight

pestle strokes. The homogenate was filtered through a 70 μm cell strainer, and 100 μL of nuclei

was incubated with 11 μL of activated ConA beads per reaction. Targeted MNase cleavage was

directed by antibodies listed in Supplementary Data 2. DNA was purified using the Zymo ChIP

DNA Clean & Concentrator Kit (#D5205), and libraries were prepared with the EpiCypher

CUTANA Library Preparation Kit (#14-1001). The libraries were pooled and sequenced in a

single lane using paired-end 150-bp reads on an Illumina Novaseq X Plus platform.

2.8 Bioinformatics Analysis of CUT&RUN Data

Raw paired-end reads were trimmed and filtered for adapter sequences and low-quality bases

using Fastp (Chen et al., 2018). Cleaned reads were aligned to the *Drosophila*

melanogaster BDGP6.46 reference genome (annotation release 113) using Bowtie2 (Langmead

and Salzberg, 2012) with end-to-end, sensitive parameters. The resulting BAM files were sorted

with SAMtools and duplicate reads removed with Picard MarkDuplicates. Peak calling was

performed using MACS3 (Zhang et al., 2008) with broad peak settings. The resulting pileup

bedGraph files were normalized to counts per million (CPM) based on the total number of

mapped reads and converted to BigWig format using bedGraphToBigWig from the UCSC

Genome Toolkit. Normalized BigWig files were used to visualize metagene and signal profiles

with deepTools (Ramírez et al., 2016), using the computeMatrix and plotProfile functions. For

peak analysis, consensus peaks (defined as peaks present in at least three out of four biological

replicates per condition) were identified using bedtools multiinter (Quinlan and Hall, 2010).

The RNAP II pause index (PI) is a gene-specific quantitative measure that assesses tau-

dependent changes in promoter-proximal pausing relative to controls. Specifically, the Ser5-P

pause index was calculated as the ratio of promoter Ser5-P RNAP II signal (transcription start

site [TSS] ± 250 bp) to gene body Ser5-P RNAP II signal (+500 bp from the TSS to −500 bp

from the transcription end site [TES]). Given the defined coordinates for promoter and gene

body regions, a minimum gene length cutoff of 1500 bp was applied. Total Ser2-P was

calculated independently as the average elongating RNAP II signal across the defined gene body

region, serving as a direct measure of productive elongation. To determine tau-dependent

differences in pause status, log2 fold changes of both the Ser5-P pause index and total Ser2-P

signal were computed between tau-expressing and control flies on a per-gene basis. Overall, the

RNAP II PI can capture relative changes in promoter-proximal RNAP II accumulation and

relative levels of the elongating RNAP II signal in the tauopathy model. RNAP II PI can be

mathematically defined as:

$$\log\left(\frac{\left(\frac{Ser5P\ Promoter}{Ser5P\ Gene\ Body}\right)^{Tau}}{\left(\frac{Ser5P\ Promoter}{Ser5P\ Gene\ Body}\right)^{Ctrl}}\right) - \log\left(\frac{(Ser2P\ Total)^{Tau}}{(Ser2P\ Total)^{Ctrl}}\right) = RNAP\ II\ Pause\ Index\ (PI)$$

3. Results

3.1 Tau^{R406W} induces progressive, age-related neurodegeneration in a *Drosophila* tauopathy
model.

The tau R406W mutation, linked to FTDP-17 (Hutton, 2000; Nakamura et al., 2019), occurs at a

conserved residue of exon 13 within the C-terminal domain, affecting all tau isoforms (Figure

1A) (Hong et al., 1998). Importantly, this mutant form of tau has structural and biochemical

properties similar to those found in AD and FTDP-17 patients (Hutton, 2000). Tau R406W

promotes hyperphosphorylation of tau, impairs its ability to bind microtubules, and increases

neuronal vulnerability to apoptotic stress, thereby contributing to neuronal cell loss observed in

tauopathy (Krishnamurthy and Johnson, 2004). In mouse models, tau R406W causes tau

hyperphosphorylation and filament aggregation, disrupting axonal transport in affected neurons

(Tatebayashi et al., 2002; Zhang et al., 2004; Ikeda et al., 2005). In *Drosophila*, pan-neuronal

expression of transgenic human tau R406W (*elav>htau^{R406W}*) (Supplementary Figure 1A-C)

reproduces key features of human tauopathies, including DNA damage (Frost et al., 2014),

impaired locomotion (Merlo et al., 2014), and neuronal cell death (Wittmann et al., 2001). Using

optic neutralization of the fly eye, we demonstrated that pan-neuronal expression of toxic

htau^{R406W} results in significant photoreceptor neuron loss starting in young, 10-day-old flies,

which markedly worsens by middle age, 20 days after eclosion (Figure 1B, C, Supplementary

Figure 1D). In contrast, neuronal expression of wild-type human tau (WT-htau) does not cause

retinal degeneration even in older flies, 40 days post-eclosion (Supplementary Figure 1E).

Notably, no rhabdomere loss was observed in 2-day-old early adult htau^{R406W} transgenic flies,

indicating that tau-induced neurodegeneration is age-dependent and not due to retinal

developmental defects. Similarly, progressive adult-onset brain vacuolization, a well-established

marker of neurodegeneration in flies, has also been reported in this model (Wittmann et al.,

2001).

To further assess age-related neurodegeneration in tau transgenic flies, we examined behavioral

indicators of neuronal function. We carried out longitudinal climbing assays on htau^{R406W}

transgenic flies and age-matched controls (no tau expression) to assess locomotion. Control flies

maintained a high, consistent climbing score across different ages. In contrast, tau^{R406W}

transgenic flies showed a 25% decrease in climbing ability at day 2 post-eclosion compared to

controls (Figure 1D). This early-adulthood climbing-deficit phenotype suggests that htau^{R406W}

expression during development impairs neuronal function without affecting overall organization

or survival. By day 10, htau^{R406W} transgenic flies demonstrated a 50% reduction in climbing

ability relative to controls, with this decline progressing to a 70% reduction by day 20 (Figure

1D). The htau^{R406W} transgenic line, hereafter referred to as tau transgenic flies, exhibits

quantifiable neurodegeneration early in life—before significant survival declines—making it an

ideal model for studying mechanisms of tau-induced neuropathology (Wittmann et al., 2001).

Additionally, they mimic the progressive, age-related neurodegenerative features seen in human

tauopathies, highlighting their relevance as a *Drosophila* model of dementia (Creekmore et al.,

2024).

3.2 Toxic tau expression drives dysregulation of neuronal, immune response, and cell cycle genes.

Accumulation of tau in *Drosophila* and mouse models, as well as human brain tissue, leads to

global chromatin relaxation and altered gene expression profiles, contributing to the development

of neurodegenerative symptoms (Frost et al., 2014). Comprehensive transcriptomic analyses of

human post-mortem brain tissue from individuals with AD and related tauopathies reveal distinct

gene expression changes that correlate with cognitive and physical impairments (Mostafavi et al.,

2018; Wan et al., 2020). Similarly, transcriptome profiling across various tau transgenic mouse

models has revealed consistent transcriptomic alterations, including decreased expression of

genes encoding synaptic proteins and increased expression of immune-response genes (Gjoneska

et al., 2015; Kim et al., 2019). Thus, we sought to determine whether the tau-associated

neurodegeneration observed in our *Drosophila* tauopathy model was accompanied by significant

and clinically relevant alterations in canonical gene expression.

We conducted longitudinal RNA-seq using RNA extracted from the heads of control and tau

transgenic flies at 2, 10, and 20 days post-eclosion (Supplementary Data 1). Three independent

biological replicates, each comprising 20 male flies, were analyzed for each genotype and time

point, except for a single discordant day 2 tauopathy fly sample, which was excluded from

further analysis (Supplementary Figure 2A, B). Principal component analysis (PCA) of the

normalized transcript counts revealed that age accounted for 51% of the total variance across all

samples, while tau expression contributed 18% (Supplementary Figure 2A). Overall, the greatest

variability in gene expression was observed among day 2 samples, likely reflecting dynamic

transcriptional changes associated with the transition to adulthood. In contrast, samples from 10-

and 20-day-old flies showed distinct genotype-specific clustering, indicating that tau expression

is the primary driver of transcriptomic differences at these later stages (Supplementary Figure

2B).

Longitudinal transcriptomic analysis revealed the greatest number of perturbations at day 10,

with 1,640 differentially expressed genes (DEGs) identified in tau-expressing flies compared to

controls (Figure 2A). In contrast, 704 and 805 DEGs were identified at days 2 and 20,

respectively (Figure 2B, C). Of all DEGs, 364 were consistently dysregulated across all three

age-time points (Figure 2D), and these were enriched for gene ontology (GO) terms related to

synaptic transmission and nervous system processes (Supplementary Data 1). This group

includes 9 of the 10 genes encoding members of the synaptic nicotinic acetylcholine receptor

(*nAChR*), a ligand-gated ion channel that regulates neuronal membrane potential. Decreases in

human homologues of acetylcholine transporters and receptors yield cholinergic neuronal

dysfunction in aging and AD (Nyakas et al., 2011). Consistently downregulated genes also

included *Syt1* and *Tomosyn,* whose human orthologs, SYT1 and STXBP5, are significantly

downregulated in postmortem AD brain tissue (Davidsson and Blennow, 1998; Canchi et al.,

2019). The consistent, age-independent dysregulation of this gene subset suggests that toxic tau

expression intrinsically alters the expression of key neuronal genes.

Notably, gene expression progressively changed with age in 10- and 20-day-old control flies

(Figure 2E). Importantly, the genes showing age-dependent expression changes in controls

exhibited exacerbated dysregulation in tauopathy flies, emerging as early as 2 days of age and

intensified markedly with age. These findings indicate that age-associated transcriptional shifts

are accelerated in a pathological tau-dependent manner, highlighting the aging cellular

environment as a key driver of tau-induced pathology and underscoring the profound impact of

age on disease-associated expression signatures.

Given the extensive neuronal cell death observed in 20-day-old tau transgenic flies, we focused

our subsequent transcriptomic analysis on 10-day-old flies to reduce the confounding effects of

neuronal cell loss on gene expression profiles (Figure 1B and Wittmann et al., 2001). At this age

time point, 17.4% of all expressed genes were differentially expressed in tau transgenic flies,

with 853 genes being downregulated and 787 genes upregulated (p-adj < 0.05, |FC| > 1.5)

(Figure 2B, Supplementary Data 1). We validated the RNA-seq findings using quantitative PCR

(qPCR) on independent replicates of 10-day-old control and tau transgenic flies, targeting 10

genes identified as significantly down- or up-regulated in the transcriptomic dataset

(Supplementary Figure 2C, D).

Consistent with previous transcriptomic studies in tau transgenic models (Gjoneska et al., 2015;

Mangleburg et al., 2020), GO term analysis of downregulated genes revealed enrichment for

categories related to synaptic organization and neuronal function, including nervous system

processes, chemical synaptic transmission, and regulation of membrane potential (Figure 2F).

Among these, *hiw*, a gene encoding a RING E3 ubiquitin ligase playing an evolutionarily

conserved role in axonal development, was notably downregulated. Its human ortholog,

MYCBP2, has been linked to autism-like neurological symptoms when mutated (AlAbdi et al.,

2023). Another gene, *Ten-a*, was also downregulated in 10-day-old tau-expressing flies, and,

notably, a potential AD-related signal has been observed near the human TENM3 ortholog,

associated with amyloidosis and tau pathology in AD cohorts (Deming et al., 2018).

Additionally, *Gad1,* which encodes the enzyme responsible for synthesizing the inhibitory

neurotransmitter gamma-aminobutyric acid (GABA), was downregulated in our tauopathy

model. This mirrors the findings in chemically inactivated neurons, the aging human frontal

cortex, and the hippocampus of AD patients, suggesting that reduced neuronal inhibition is a

shared feature of aging and AD in both flies and humans (Blalock et al., 2004; Lu et al., 2004;

Gleichmann et al., 2012).

In 10-day-old tau transgenic flies, upregulated genes were significantly enriched for GO terms

associated with anabolic biological processes, such as small molecule biosynthesis and

cytoplasmic translation, as well as various catabolic pathways, including amino acid and

carbohydrate metabolism (Figure 2G). Notably, several cytochrome P450 genes - *Cyt-b5-r*,

Cyp4d1, *Cyp4e2*, *Cyp9b1*, and *Cyp6a17* - were upregulated, aligning with human studies that

report elevated levels of cytochrome P450 components in the cerebrospinal fluid and plasma in

antemortem AD patients over healthy controls (Borkowski et al., 2021). Additionally, genes

encoding proteins involved in immune response, particularly those from the immune deficiency

(imd) and Toll signaling pathways, were markedly upregulated in tau transgenic flies. This

robust activation of immune-related genes underscores chronic inflammation as a conserved

feature of tauopathy in both flies and humans (Langworth-Green et al., 2023).

An increased body of evidence suggests that tau contributes to neurodegeneration by aberrantly

reactivating the cell cycle in post-mitotic neurons, ultimately leading to apoptosis (Khurana et

al., 2006; Frost et al., 2014; Zhao et al., 2021a). Supporting this mechanism, our transcriptomic

analysis revealed upregulation of genes encoding proteins involved in DNA replication and cell

cycle progression, including proliferating cell nuclear antigen (*PCNA*), *Nnf1b*, *Trithorax-like*

(*Trl*), *Origin recognition complex subunit 1* (*Orc1*), and Minichromosome maintenance 2

(*Mcm2*). We also observed increased expression of *Nucleophosmin* (*Nph*), a gene that typically

protects neurons but, when overexpressed, can promote cell cycle re-entry and apoptosis (Pfister

and D'Mello, 2015). Conversely, gene expression of Wee1 kinase (*Wee1*), a key regulator of the

G2/M checkpoint, was reduced, further supporting dysregulation of cell cycle control (Esposito

et al., 2021).

Collectively, these findings highlight the similarities between gene expression changes in the

Drosophila tauopathy model and those observed in human Alzheimer's disease. They also help

disentangle the distinct and overlapping contributions of aging and tau pathology to

transcriptomic alterations. Importantly, the gene expression changes identified in the 10-day-old

tau transgenic flies reveal early transcriptional disruptions associated with tau toxicity. These

include reduced synaptic signaling, increased immune response, and aberrant cell cycle

activation, all of which are key features of progressive tau-related neurodegeneration.

3.3 Long genes with long introns are preferentially and significantly downregulated in tau

transgenic *Drosophila*.

We previously demonstrated that long, highly expressed genes encoding proteins with neuronal

function tend to decrease expression in aging *Drosophila* photoreceptor neurons; a decline that

correlates with age-related visual impairment (Hall et al., 2017). Similar associations between

gene length, transcriptional output, and cellular function have been reported across various

tissues and organisms in both aging and neurodegenerative disease contexts (Barbash and

Sakmar, 2017; Lopes et al., 2021; Stoeger et al., 2022; Debès et al., 2023; Gyenis et al., 2023).

To explore whether similar features characterize genes dysregulated in our tau transgenic flies,

we analyzed transcriptomic data based on conserved genomic attributes and expression levels. In

10-day-old tau transgenic flies, downregulated genes were significantly longer than upregulated

or unchanged genes (Figure 3A). Importantly, average transcript abundance did not differ

significantly between downregulated genes and other differential expression groups

(Supplementary Figure 3A). GO term analysis of long downregulated genes (>10kb) revealed

enrichment for categories associated with neuronal structure and function, including synapse

organization and signaling (Figure 3B). Since neuronal genes are generally longer than those

expressed in other cell types (Zylka et al., 2015), we controlled for potential bias by focusing on

genes associated with GO0050877 (Nervous System Process), a highly enriched and broadly

encompassing neuronal GO term among downregulated genes (Figure 2C). Even within this

subset, downregulated genes were significantly longer than genes in other differential expression

categories (Figure 3C). In contrast, long genes that did not change expression in tau transgenic

flies were associated with cell morphogenesis and developmental processes (Figure 3D).

Gene length is shaped by both exon and intron architecture. While exons are short and highly

conserved, introns vary widely in length and can impact transcription efficiency (Carmel et al.,

2007; Yenerall and Zhou, 2012). In our tauopathy model, long genes that were significantly

downregulated contained more exons than genes with unchanged expression and approximately

twice as many as those in the upregulated gene group (Figure 3E). However, the total length of

their introns was four times that of genes with unchanged expression, averaging approximately

22 kb, compared with approximately 5 kb for introns in genes with unchanged expression

(Figure 3F). Long intron splicing deficits in the tauopathy model may contribute to gene

dysregulation; however, at the transcript level, there is no change in the expression of key

splicing factors (Supplementary Figure 3B). Overall, the observed difference in intron length

largely explains the increased total gene length observed in downregulated genes, suggesting that

this genomic feature may play a critical role in regulating proper gene expression in tauopathy

flies.

3.4 Increased promoter-proximal RNA polymerase II pausing is associated with long downregulated genes.

To focus our investigation of tau-dependent dysregulation of gene expression, we assessed

transcription dynamics using genome-wide profiling of RNAP II. After transcription initiation,

RNAP II typically pauses near the promoter, and its timely release into productive elongation is a

key regulatory step in fine-tuning gene expression. To map the RNAP II distribution in tau

transgenic flies, we assayed 10-day-old tau and age-matched control flies using the CUT&RUN

technique. This approach, combined with antibodies specific for Ser5- and Ser2-phosphorylated

(Ser5- and Ser2-P) RNAP II, markers of initiating and elongating forms of the polymerase,

respectively (Bozukova et al., 2022), enabled us to assess RNAP II dynamics with high

resolution.

CUT&RUN generated 3-6 million uniquely mapped fragments across four biological replicates

per genotype and antibody. Principal component analysis revealed that the primary source of

variance was RNAP II phosphorylation state (PC1) and tau expression (PC2) (Supplementary

Figure 4A). In control flies, metaplot analysis revealed a sharp RNAP II peak around the TSS,

followed by a plateau across gene bodies and a decrease at the transcription end site (TES)

(Figure 4A, first column, gray line), consistent with a canonical RNAP II pattern across actively

transcribed genes (Bozukova et al., 2022). In contrast, tau transgenic flies showed elevated levels

of initiating RNAP II at the TSS and reduced levels of elongating RNAP II around the promoter

and throughout the gene body (Figure 4A, first column, yellow line). Stratifying RNAP II

profiles by gene expression revealed that both up- and downregulated genes in tau transgenic

flies exhibited increased initiating RNAP II occupancy at the TSS (Figure 4A, top row). For

upregulated genes, elongating RNAP II levels were comparable between tau-transgenic and

control flies, suggesting that increased transcript levels may result from enhanced transcription

initiation or increased RNA stability (Figure 4D, bottom row). Conversely, downregulated genes

exhibited a marked depletion of elongating RNAP II across gene bodies, consistent with reduced

transcriptional output.

To more specifically quantify pause proclivity in the tauopathy model, we calculated the RNAP

II Pause Index (PI), an initiation-to-elongation transition factor, for each gene. Building on

previously described RNAP II pause-release ratios (Akhtar et al., 2021), we integrated our

phospho-specific RNAP II dataset by defining the PI as the ratio of promoter-enriched Ser5-P

RNAP II to total gene body Ser2-P RNAP II occupancy for each gene in tauopathy versus

control flies (Figure 4B). Using this RNAP II PI as an unbiased, gene-level measurement of

promoter-proximal pausing, we uncovered a strong relationship between polymerase pausing,

gene expression changes, and gene length (Figure 4C, Supplementary Figure 4B). Specifically,

genes exhibiting significantly increased promoter-proximal pausing were long and

downregulated in tau-transgenic flies. Further, within the most-paused gene cohort, those with

the highest average RNAP II PI, we identified several long, downregulated neuronal genes,

including *Tomosyn* and *Frq1*, that exemplify the canonical pattern of increased promoter

proximal RNAP II pausing (Figure 4D, E).

These findings reveal that RNAP II dynamics are specifically disrupted in the tauopathy model.

Notably, downregulated genes in tau transgenic flies are marked by the concomitant

accumulation of initiating RNAP II and depletion of elongating RNAP II. Taken together, these

RNAP II profiles suggest that exacerbated promoter-proximal polymerase pausing is a hallmark

of long, downregulated genes in the tauopathy model. This promoter-proximal polymerase

stalling may underlie failure of critical transcriptional processes, particularly the transition from

initiation to elongation, as a potential mechanism driving transcription stress in tauopathy.

4. Discussion

Gene dysregulation is a central hallmark of tauopathies, including Alzheimer's disease, and underpins a range of pathological symptoms. Despite growing recognition of its importance, our understanding of the multilayered regulation of gene expression, encompassing transcriptional, epigenetic, and post-transcriptional mechanisms, in neurodegenerative disease remains limited, particularly in the context of aging (Miller et al., 2008, 2010; Wan et al., 2020; Tan et al., 2021). In this study, we used a well-established *Drosophila* tauopathy model with neuronal expression of toxic human tau^{R406W}. Previously, the age-dependent transcriptome and proteome of this tauopathy model were characterized in a comparative analysis to identify intersectional trends in RNA and protein expression (Mangleburg et al., 2020). Here, we further interrogated the ubiquitous and age-dependent transcriptional alterations driven by tau accumulation and identified conserved genomic features of downregulated genes to elucidate the underlying molecular mechanism driving gene dysregulation.

Tau-expressing flies exhibited preferential downregulation of genes required for maintaining neuronal structure and function, including those involved in learning, memory, and locomotion. Concurrently, we observed upregulation of genes associated with stress and immune responses, metabolism, and cell cycle regulation. This tau-dependent gene-dysregulation profile recapitulates age-related changes in expression. However, tau markedly accelerates and amplifies age-dependent transcriptional stress, underscoring how aging and disease converge to shape pathological gene expression signatures. These disrupted gene expression patterns also mirror findings from other tau transgenic models, including aberrant cell cycle re-entry of post-mitotic neurons, a process widely implicated as a key driver of neurodegeneration in tauopathies and AD (Khurana et al., 2006; Zhao et al., 2021b).

Characterization of distinct genomic trends among differentially expressed genes in the tauopathy model revealed that downregulated genes were significantly longer and had a four-fold increase in cumulative intron length compared to non-differentially expressed genes. While neuronal transcripts are generally longer than those of other cell types (Zylka et al., 2015), the subset of downregulated neuronal genes in our dataset was notably longer than other neuron-specific genes that were not differentially expressed. These findings align with a growing body of research indicating an inverse relationship between gene length and transcriptional output in

the context of aging and neurodegenerative disease (Zylka et al., 2015; Hall et al., 2017; Gyenis et al., 2023).

Further analysis revealed that long, downregulated genes in tau-expressing flies showed reduced elongating RNAP II occupancy across gene bodies. These downregulated genes also showed an accumulation of the initiating form of RNAP II just downstream of the transcription start site, suggesting a failure to release promoter-proximal paused polymerase into active elongation. On active genes, the transition of primed RNAP II into elongation is inhibited by negative regulators NELF and DSIF, promoted by P-TEFb-mediated phosphorylation, and maintained by other key histone modifiers and RNA-binding proteins (Jonkers and Lis, 2015). At the transcript level, we observed no significant changes in the expression of these critical elongation factors in the tauopathy fly model (data not shown). However, recent studies have shown that transcription factors, splicing factors, and RNA-binding proteins can co-localize with pathogenic tau tangles in the cytoplasm (Hsieh et al., 2019; Kavanagh et al., 2022), potentially limiting their nuclear availability and regulatory function.

Notably, recent work in *Drosophila* shows that the nuclear exon junction core complex (pre-EJC) associates with promoter regions through interactions with the initiating, Ser5-P RNAP II, mediated by nascent RNA transcripts, thereby reinforcing promoter-proximal pausing. This effect is particularly pronounced at genes containing long introns (Akhtar et al., 2019). In this context, our findings further underscore, in particular, gene and intron length as critical determinants of tau-induced transcriptional stress, suggesting that genes with these attributes are especially susceptible to transcriptional dysregulation.

Gene dysregulation is a defining molecular hallmark of age-related diseases; yet, how these regulatory pathways become disrupted in specific disease contexts remains a central question in neurodegenerative disease biology. Our data reveal distinct transcriptional dynamics and genomic features associated with downregulated transcripts in tauopathy, providing compelling evidence that promoter-proximal RNAP II stalling is a hallmark of long neuronal genes. These findings underscore the importance of studying gene regulatory mechanisms within disease-relevant contexts and highlight a pathological shift in transcriptional control in tauopathy. Future studies that elucidate the key factors in tau pathology that drive promoter-proximal RNAP II stalling and further clarify the connection between genomic features, such as gene and intron

length, and RNAP II dynamics at vulnerable neuronal genes will help unravel the complexity of

transcriptional regulation in age-related tauopathy. Furthermore, given the close interplay

between genomic architecture, nascent RNA modifications, and chromatin structures in

transcription regulation, investigating how these co-transcriptional elements collectively shape

gene expression in neurodegeneration will be of significant interest.

Figure Legends

Figure 1. (A) Schematic of the human tau transgene isoform showing defined domains and the

location of the R406W mutation. **(B)** Schematic illustrating the ageing time points used for

phenotypic analysis. **(C)** Boxplot quantification of retinal degeneration in aged control

(*elav>lacZ*) and tau (*elav>htau^{R406W}*) flies. Rhabdomeres of 10 individual ommatidia per fly

(n=10 male flies) were assessed. Retinal degeneration scores were calculated blindly as the ratio

of the number of lost rhabdomeres to the total number of rhabdomeres, indicating photoreceptor

cell death. **(D)** Boxplots showing the climbing ability of aged control (*Cyo/ htau^{R406W}*) and tau

(*elav>htau^{R406W}*) flies. 6-12 replicates of 25 male flies were assessed for each genotype. Student's

t-test was used to determine statistical significance between genotypes at each time point (**** p

< 0.001, **$p < 0.01$, *$p < 0.05$, NS/ns not significant).

Supplementary Figure 1. (A) A diagram illustrating a genetic approach to create a *Drosophila*

tauopathy model using pan-neuronal *elav-Gal4*-driven expression of the toxic hTau^{R406W}.

Relative htau RNA **(B)** and protein **(C)** levels in 10-day-old transgenic flies. n=3 replicates of 25

male flies per genotype. **(D)** Left: Schematic of the *Drosophila* compound eye showing a single

ommatidium enlarged to display photoreceptor neurons (R1-7) with rhabdomeres (red circles).

Right: Representative images showing the loss of rhabdomeres in photoreceptor neurons

(indicated by yellow arrowheads) in tau (*elav>htau^{R406W}*) flies. **(E)** Boxplot quantification of

retinal degeneration observed in aged flies expressing WT-htau. Student's t-test was used to

determine statistical significance between genotypes at each time point (NS/ns not significant).

Figure 2. (A) Volcano plot showing differentially expressed genes (DEGs) between tau and

control flies at day 2, **(B)** day 10, and **(C)** day 20, post-eclosion. DEGs are identified using

DESeq2 (adjusted p value < 0.05, |FC| > 1.5). **(D)** Venn diagram illustrating the overlap between

downregulated and upregulated genes across all aging timepoints. **(E)** Heatmap showing trends

in expression Z-scores for all genes across all samples and aging timepoints. **(F)** Dot plot

displaying the 20 most significantly enriched biological process gene ontology (GO) terms for

downregulated genes in day 10 tau transgenics. **(G)** Dot plot illustrating the 15 most significantly

enriched biological process GO terms for upregulated genes in day 10 tau transgenics.

Supplementary Figure 2. (A) Principal component analysis (PCA) of longitudinal RNA-

sequencing samples based on normalized counts per million. **(B)** Hierarchically clustered

heatmap of Pearson correlation coefficients calculated from normalized RNA-seq expression

values across all samples. Bar plots displaying the relative transcript levels of genes assessed by

qPCR that were identified as downregulated **(C)** or upregulated **(D)** in tau transgenic flies by

RNA sequencing. n=3-4 replicates of 20 male flies per genotype.

Figure 3: (A) Violin and boxplot of differentially expressed gene (DEG) categories showing the

total length of all genes. **(B)** Dot plot displaying the 15 most enriched GO biological process

terms associated with long, downregulated genes. **(C)** Boxplot showing the length of genes

within 'GO0050877: Nervous System Process' by DEG categories. **(D)** Dot plot displaying the

15 most enriched GO biological process terms associated with long non-DEGs. Violin and

boxplot of differentially expressed gene (DEG) categories showing **(E)** number of exons, and **(F)**

total intron length. For differential expression data, DOWN (downregulated), NO (no change),

and UP (upregulated) expression in tau transgenic flies compared to controls at day 10 post-

eclosion. Long genes are defined as those > 10kb. ANOVA followed by TukeyHSD post-hoc

test was used to evaluate statistical significance between DEG groups (**** $p < 0.001$, * $p <$

0.05, NS/ns not significant).

Supplementary Figure 3. (A) Violin and boxplot illustrating the expression of all long genes

based on DEG categories. Differential expression data are indicated as DOWN (downregulated),

NO (no change), and UP (upregulated) expression in tau transgenic flies compared to controls at

day 10 post-eclosion. Long genes are defined as > 10kb. ANOVA followed by TukeyHSD post-

hoc test was used to determine statistical significance between DEG groups (**** $p < 0.001$, ***

p < 0.005, ** p < 0.01, *p< 0.05, NS/ns, not significant). **(B)** Dot plot displaying the 15 most

enriched GO terms among all non-DEGs.

Figure 4. (A) Metaplots of CPM (counts per million)-normalized Ser2- and Ser5-phosphorylated

RNAP II CUT&RUN signals over gene bodies of all expressed protein-coding genes, all up- and

downregulated genes for control (gray) and tau transgenic flies (yellow). TSS, transcription start

site; TES, transcription end site. **(B)** Calculations for determining RNAP II Pause Index (PI). PI

values greater than 0 indicate increased promoter-proximal pausing for a given gene in the

tauopathy model. **(C)** Dot plot showing RNAP II pausing index (PI) values, gene expression, and

gene length. Genes were ranked by increasing PI score and grouped into 20 equal-sized bins.

Bin-wise means of PI value, gene expression, and gene length were plotted to visualize global

trends. **(D)** Genome browser inspection of RNAP II profiles in the Integrated Genomic Viewer

(IGV) depicting a representative long, downregulated neuronal gene, *Tomosyn*, from the highest

RNAP II PI bin. RNAP II density represents the average of four independent biological

replicates per genotype and RNAP II phosphor-isoform.

Supplementary Figure 4. (A) PCA plot of RNAP II CUT&RUN samples based on consensus

peak counts. The dot shape indicates genotype, and the dot color indicates antibody target. **(B)**

Dot plot showing the correlation between RNAP II PI and gene expression for all expressed

genes with an adjusted p-value <0.05. Genome browser inspection of RNAP II profiles in the

Integrated Genomic Viewer (IGV) depicting representative long, downregulated neuronal genes,

Frq1 **(C)** and *nAChRalpha6* **(D)**, from the highest RNAP II PI bin. RNAP II density represents

the average of four independent biological replicates per genotype and RNAP II phosphor-

isoform.

Statements

Conflict Of Interest

The authors declare that the research was conducted in the absence of any commercial or
financial relationships that could be construed as a potential conflict of interest.

Author Contributions

KSC: Data curation, Formal Analysis, Investigation, Methodology, Visualization, Writing – original draft, review, and editing. NG: Formal analysis, Visualization, Writing–review. DF: Validation, Writing–review. GL: Validation, Writing–review. HH: Conceptualization, Funding acquisition, Resources, Supervision, Writing-review, and editing.

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Data Availability Statement

The raw/processed RNA-seq and CUT&RUN files are deposited in a repository and will be

released upon acceptance or, if submitted concurrently with the publication, upon publication.

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(A)

N terminus | Proline-rich domain | Microtubule binding domain | C terminus
0N | | 4R | *
R406W

(B)

Ctrl

Tau

Aging

Day 2: Early Adult

Day 10: Young Adult

Day 20: Middle-Aged Adult

Neurodegenerative Phenotyping Assays

(C)

Retinal Degeneration (%)

D2 — NS. — Ctrl Tau

D10 — ** — Ctrl Tau

D20 — ** — Ctrl Tau

(D)

Climbing Score

D2 — **** — Ctrl Tau

D10 — **** — Ctrl Tau

D20 — **** — Ctrl Tau

(A)

elav > Gal4 ♀ ✕ *UAS-hTau^{R406W}* ♂

Neuronal tau expression

elav — *Gal4*   *UAS* — *hTau^{R406W}*

(B)

D10 Tau Transgene Expression

Relative Gene Expression

Ctrl   Tau

*

(C)

	Ctrl	Tau	kDa
Tau			60
Tubulin			60 / 45

(D)

R1-R8 photoreceptor neurons

ommatidium

R4 R3 R5 R2 R8 R6 R7 R1

D2   D10   D20

Ctrl

hTau(R406W)

(E)

Retinal Degeneration (%)

D10   D20   D30   D40

NS.   ns   ns   ns

Ctrl hTau(WT)   Ctrl hTau(WT)   Ctrl hTau(WT)   Ctrl hTau(WT)

(A)

(B)

(A)

All Genes   Up   Down

Initiating RNAP II Ser5-P

Elongating RNAP II Ser2-P

(B)

$$\log\left(\frac{\left(\frac{Ser5P\ Promoter}{Ser5P\ Gene\ Body}\right)^{Tau}}{\left(\frac{Ser5P\ Promoter}{Ser5P\ Gene\ Body}\right)^{Ctrl}}\right) - \log\left(\frac{(Ser2P\ Total)^{Tau}}{(Ser2P\ Total)^{Ctrl}}\right) = RNAP\ II\ Pause\ Index\ (PI)$$

(C)

Slope = -0.658
Intercept = 0.303
R² = 0.881

Mean gene length

(D)

Tomosyn (chrX:12,631,305-12,653,620)

(A)

(B)

R = -0.36 p < 2.2e-16

(C) Frq1 (chrX:18,153,065-18,181,260)

(D) nAChRalpha6 (chr2L:9,791,260-9,888,193)