Somatic mosaicism in amyotrophic lateral sclerosis and frontotemporal dementia identifies focal mutations associated with widespread degeneration

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Abstract
Although mutations in many genes cause familial amyotrophic lateral sclerosis and
frontotemporal dementia, most cases are sporadic (sALS and sFTD) with unclear etiology. We
tested whether somatic mutations contribute to sALS and sFTD by deep targeted sequencing of
88 neurodegeneration-related genes in postmortem brain and spinal cord samples from 399
sporadic cases and 144 controls. Predicted deleterious somatic variants in ALS/FTD genes were
observed in 2.1% of sporadic cases lacking deleterious germline variants. These variants
occurred at very low allele fractions (typically <2%) and were often focal and enriched in
disease-affected regions. Analysis of bulk RNA-seq data from an additional cohort identified
deleterious somatic variants in *DYNC1H1* and *LMNA*, genes associated with pediatric motor
neuron degeneration. Targeted long-read sequencing further identified one sFTD case with *de*
novo somatic *C9orf72* repeat expansions. Together, these findings suggest that rare, focal
somatic variants can contribute to sALS and sFTD and drive widespread neurodegeneration.
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Introduction

Amyotrophic lateral sclerosis (ALS), a disease in which premature loss of upper and lower motor neurons (UMNs and LMNs) leads to fatal paralysis, shows clinical, genetic, and pathological overlap with frontotemporal dementia (FTD), a neurodegenerative disorder characterized by behavioral, language, and memory dysfunction[1]. 5-22% of individuals with ALS develop FTD, and ~15% of those with FTD eventually develop ALS[2]. ALS and FTD also share common pathology, with cytoplasmic inclusions of TAR DNA binding protein (TDP-43) found in almost all ALS brains and in half of FTD brains[3,4]. ALS typically begins focally and spreads regionally as the disease progresses[5,6], although whether degeneration begins in UMNs, LMNs, or both simultaneously has remained controversial[7,8], with some studies suggesting that focality can manifest independently in UMNs and LMNs[5,9]. TDP-43 pathology also follows stereotypical patterns in ALS and FTD brains[9-11], thought to reflect focal onset and intercellular transmission of TDP-43 inclusions in a prion-like manner, as shown in cell and animal models[12-18].

Whereas over 30 genes are implicated in ALS and FTD[19], most causative genes are linked to familial ALS (fALS) and FTD (fFTD), while 90-95% of cases are sporadic ALS (sALS) and FTD (sFTD) without a family history[20]. The focal onset of ALS and FTD, their stereotypical spread, and the increased risk in smokers[21] have raised interest in potential roles of somatic mosaic mutations in the pathogenesis of ALS and FTD[22]. Somatic mutations are increasingly recognized as prevalent in normal-appearing tissues, but somatic mutations responsible for neurological conditions are often limited to the central nervous system (CNS)[23], and hence undetectable through DNA sequencing of non-CNS tissues. Recent studies have evaluated the contributions of somatic mutation to Alzheimer's and Parkinson's disease directly using postmortem brain tissues[24].

In this study, we assessed potential contributions of somatic variants—distinguished by their variant allele frequencies (VAFs)—to sALS and sFTD using deep sequencing of a panel of neurodegeneration/dementia-associated genes on postmortem tissues of various brain regions and spinal cords from 399 unique sALS and sFTD cases. Our study identified novel predicted deleterious somatic variants in 2.1% of the sALS and sFTD cases without pathogenic or predicted deleterious germline variants. Protein-altering (missense/nonsense/frameshift) somatic variants showed enrichment in sALS and sFTD cases and in disease-affected brain regions, supporting roles in disease pathogenesis. Regional analysis revealed focality of predicted deleterious somatic variants in primary motor cortex and spinal cord, supporting independent disease initiation in UMNs and LMNs, but also strongly supporting models of ALS and FTD in which the disease spreads beyond a relatively confined region containing a somatic variant. Complementary analyses of bulk RNA-seq and targeted long-read sequencing of brain and spinal cord tissues further revealed somatic variants in genes not previously associated with ALS or FTD, as well as a *de novo* somatic *C9orf72* repeat expansion. Together, our study opens new avenues for understanding the etiology of sporadic ALS and FTD.

Results

Deep targeted sequencing of neurodegenerative genes in sALS and sFTD brains

To directly detect somatic variants in sALS and sFTD brains, we obtained post-mortem frozen tissues of several brain regions and spinal cords from individuals diagnosed with sALS or sFTD, as well as from age-matched controls through the Massachusetts Alzheimer's Disease Research Center, Oxford Brain Bank, and Target ALS Foundation (Fig. 1a and Supplementary Table 1). Additional brain tissues from ALS, FTD and control cases, without a record of family history but

94   with an age of death above 45 years old, were also obtained from the NIH NeuroBioBank. We
95   performed molecular inversion probe (MIP)-panel sequencing[25] of 88 neurodegeneration-
96   associated genes at ~1,800× deduplicated depth across 1,787 samples from 291 ALS, 117 FTD,
97   and 144 neurotypical control individuals (Fig. 1a,b, Supplementary Fig. 1, Supplementary Table
98   2, and Supplementary Note).
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Pathogenic germline variants in sALS and sFTD cases
We first identified pathogenic germline single-nucleotide variants (SNVs) and short insertions
and deletions (indels) using GATK followed by stringent filtering (Fig. 1d), with functional
annotation and deleteriousness prediction performed using ANNOVAR[26] and multiple clinical
databases. In addition, *C9orf72* repeat expansions, the most common inherited cause of ALS and
FTD[27,28], were genotyped by a repeat-primed PCR assay (Supplementary Fig. 2). Overall, 20.6%
(60/291) of ALS, 25.6% (30/117) of FTD and 0.7% (1/144) of control cases carried *C9orf72*
repeat expansions or pathogenic germline mutations in ALS and FTD genes (Fig. 2a and
Supplementary Tables 3 and 4). Missense variants represented the most prevalent variant type
(Fig. 2b). *C9orf72* was the most frequently mutated gene, followed by *SOD1* in ALS, and *GRN*
and *MAPT* in FTD (Fig. 2c,d). The overall fractions of *C9orf72* repeat expansion carriers (10.6%
in ALS; 12.0% in FTD) were slightly higher than previously reported in sporadic cases, but still
lower than in familial cases[29-31].
In addition, predicted deleterious germline variants in dominant ALS/FTD genes were
found in an additional 14.1% of ALS, 18.8% of FTD, and 4.9% of control cases with significant
enrichment observed in both ALS (OR = 3.20, 95% CI: 1.37-8.69, $P = 3.2 \times 10^{-3}$) and FTD (OR
= 4.51, 95% CI: 1.77-13.01, $P = 5.5 \times 10^{-4}$) cases (Fig. 2a and Supplementary Tables 3 and 4).
While most predicted deleterious variants were non-synonymous SNVs requiring functional
validation, two novel *GRN* frameshift variants (p.L46Rfs*18 and p.D250Tfs*6) identified in
FTD cases were probably pathogenic (Supplementary Table 3), since loss-of-function *GRN*
mutations are known to cause FTD in a dominant manner[32,33]. Consistent with previous reports,
we found evidence of possible oligogenic inheritance. This includes one case carrying both
C9orf72 expansion and a pathogenic variant in *ANXA11*, as well as 12 additional cases carrying
combinations of pathogenic and predicted deleterious variants (Fig. 2d and Supplementary Table
3)[34-37]. Interestingly, we also identified cross-disease—variants in ALS genes in FTD and vice
versa—highlighting potential shared genetic mechanisms (Fig. 2d).
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Identification of somatic SNVs and indels from MIP sequencing data
We developed a custom pipeline integrating RePlow[38], Mutect2[39], and Pisces[40] for calling
somatic SNVs and indels in our MIP sequencing data (Fig. 1d). We selected somatic variants
identified by at least two of the three callers (double-called variants) followed by multi-step
variant filters to remove false positive candidates. Unlike heterozygous germline variants with
VAFs around 50%, heterozygous somatic variants have VAFs less than 50%, and we only called
somatic variants with VAFs below 35%. To benchmark our pipeline, we performed spike-in
experiments by mixing two human samples from the Genome in a Bottle Consortium (GIAB) at
low VAFs and observed high sensitivity and precision with a low false positive rate across VAFs
(Extended Data Fig. 1a,b and Supplementary Note).
We applied our custom pipeline to 1,787 samples to obtain initial somatic variant
candidates. Among these, we identified and removed candidates arising from potential sample
contamination in 29 samples (Supplementary Note). The remaining call set contained 98 unique

somatic SNV and indel candidates (a total of 149 candidates, including shared ones observed in
multiple regions of the same individuals; see Supplementary Table 5). Variants with low VAFs
(<5%) were more common in disease cases than in normal controls (Extended Data Fig. 2). All
somatic candidates were validated by deep amplicon sequencing, except one for which primers
could not be designed; 34 exonic candidates were additionally assessed by droplet digital PCR
(ddPCR). Thirteen candidates were actually germline variants, consistent with known VAF
deviation in MIPs sequencing due to probe hybridization and amplicon design biases[41,42]. After
excluding these, 64 out of 85 candidates (75.2%) were validated as somatic (Supplementary
Table 6). The VAFs of validated candidates in amplicon sequencing showed a strong correlation
with their original VAFs in the MIP sequencing data (Fig. 3a). All subsequent somatic variant
analyses were conducted using only the validated candidates.
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152   Somatic variants in disease-relevant genes are enriched in ALS and FTD cases lacking
153   pathogenic germline variants
To examine the burden and potential roles of somatic variants in ALS and FTD, we focused on
cases lacking pathogenic or predicted deleterious germline variants in ALS/FTD genes
(germline-free cases), including cases with heterozygous germline variants in recessive
ALS/FTD genes (e.g., *ATP13A2*, *ALS2*) that were predicted to be deleterious. Fifty-five unique
somatic variants across the targeted regions were identified among 696, 243, and 516 samples
from 216 germline-free ALS cases, 78 germline-free FTD cases, and 144 neurotypical controls,
respectively. Most of them (78.2%, 43/55) were focal to a single tissue region and present at very
low VAFs (Fig. 3b and Extended Data Fig. 2), consistent with late-arising[43], CNS-restricted
events. Mutational signature analysis [44] revealed predominance of clock-like signatures (SBS5
and SBS1) (Extended Data Fig. 3), consistent with previous studies of somatic mutagenesis in
the normal human brain[45,46].
165   Across all targeted neurodegeneration/dementia-related genes, protein-altering somatic
variants showed clear separation between disease and control groups (Fig. 3c), with only one
such variant observed in controls compared with 15 in ALS and 7 in FTD cases. Both exonic and
protein-altering somatic variants were significantly enriched in ALS and FTD (Fig. 4a; exonic: *P*
= 0.044 and *P* = 0.0014; protein-altering: *P* = 0.019 and *P* = 0.010; linear mixed model; see
Supplementary Note), whereas intronic and non-coding somatic variants showed no enrichment.
Consistently, ratios of non-synonymous to non-coding variants were increased in ALS and FTD
at both group and individual levels (Extended Data Fig. 4), indicating disease-specific
accumulation of potentially damaging variants.
174   When restricted to ALS/FTD genes, exonic and protein-altering variants in established
ALS genes were specifically enriched in germline-free ALS cases (Fig. 4b; exonic: *P* = 0.046;
protein-altering: *P* = 0.026; linear mixed model). In contrast, germline-free FTD cases showed
no enrichment in known FTD genes but exhibited significant enrichment in other
neurodegeneration/dementia-related genes (Fig. 4a,b and Extended Data Fig. 5), consistent with
broader pathological heterogeneity of FTD, and perhaps implicating a shared genetic architecture
with other neurodegenerative diseases. No protein-altering variants in ALS/FTD genes were
observed in controls or in ALS/FTD cases carrying pathogenic germline variants (Supplementary
Tables 6 and 7).
183   Somatic variants also exhibited region-specific enrichment in disease-affected regions of
germline-free ALS and FTD cases (Fig. 4c and Extended Data Fig. 6). Across all targeted genes,
enrichment was observed in the primary motor cortex in ALS cases (Extended Data Fig. 6; all: *P*

= 0.020; exonic: P = 0.026; protein-altering: P = 0.012; linear mixed model) and in the prefrontal
cortex of FTD brains (Extended Data Fig. 6; exonic: P = 2.3 × 10^{-3}; protein-altering: P = 7.7 ×
10^{-4}; linear mixed model). In contrast, the premotor cortex—located immediately between the
primary motor cortex and prefrontal cortex—showed no enrichment for either condition. When
restricted to ALS/FTD genes, ALS cases still demonstrated the enrichment only in the primary
motor cortex (Fig. 4c; all: P = 8.3 × 10^{-3}; exonic: P = 0.060; protein-altering: P = 0.029; linear
mixed model), while FTD cases showed no significant enrichment in the prefrontal cortex,
further supporting a broader genetic heterogeneity in FTD. The spinal cord in ALS showed a
modest increase in exonic and protein-altering variants, although this analysis is limited by the
small number of control samples and wide confidence intervals (Fig. 4c). Together, these
diagnosis- and region-specific patterns suggest that functional somatic variants may contribute to
the pathogenesis of sALS and sFTD.
198
Predicted deleterious somatic variants have restricted regional distributions and are
enriched in hypodiploid cells
Pathogenicity prediction identified six predicted deleterious somatic SNVs in known ALS and
FTD genes (Supplementary Table 7), which account for 2.3% and 1.3% of germline-free ALS
and FTD cases, respectively (2.1% for overall). All variants in ALS cases were observed in
primary motor cortex or spinal cord, the most severely affected regions in ALS, emphasizing
remarkable topographic specificity. All somatic variants occurred in disease genes with dominant
inheritance when found in the germline setting, except for one sALS case with a predicted
deleterious somatic *ALS2* (p.T787R) variant identified in the spinal cord. *ALS2* is an autosomal
recessive disease gene[47,48], and this individual also carried a predicted deleterious germline *ALS2*
(p.Q24R) variant in addition to the identified somatic variant. Both *ALS2* variants were predicted
to be deleterious, consistent with a "second hit" mechanism at the cellular level in a small
proportion of spinal cord cells.
We selected four predicted deleterious somatic SNVs—*TIA1* (p.H54R), *MATR3*
(p.K594I), *ALS2* (p.T787R), and *TARDBP* (p.L248F)—for detailed analysis of regional and cell-
type distributions. Amplicon sequencing across multiple CNS regions showed that *MATR3*
(p.K594I) and *TARDBP* (p.L248F) were restricted to the primary motor cortex (Fig. 4d and
Supplementary Table 8), whereas *TIA1* (p.H54R) and *ALS2* (p.T787R) exhibited their highest
VAFs in the spinal cord (2.16% and 0.97%, respectively), where they were originally identified,
and were detected at much lower levels in other brain regions (Fig. 4d and Supplementary Table
8). All four somatic SNVs were absent from the cerebellum. The ultra-low VAFs and focal
distribution of these variants suggest that they probably arose late in development and were thus
likely CNS-restricted. Together with the enrichment of exonic and protein-altering somatic
variants in disease-affected tissue regions, these findings also support the focal onset of ALS at
the genetic level in these sporadic cases. Cells carrying damaging somatic variants could form
initial lesions, likely TDP-43 inclusions, in UMNs and LMNs, with pathology ultimately
spreading to other regions of the motor system that lack or carry exceedingly low levels of the
variant, but nonetheless show robust post-mortem pathology otherwise indistinguishable from
germline cases. Consistent with this hypothesis, the case harboring the *TARDBP* (p.L248F)
variant, showed the highest phospho-TDP-43 (pTDP-43) level in the primary motor cortex,
where the variant was detected (Extended Data Fig. 7). Quantification across seven brain regions
revealed significant regional differences in pTDP-43 levels (F = 3.20, P = 0.0099, one-way
ANOVA), with the motor cortex exhibiting significantly higher pathology compared to the

hippocampus ($P = 0.0042$), middle temporal gyrus ($P = 0.040$), and occipital cortex ($P = 0.032$)
in post hoc Tukey's HSD tests. These results support the idea that focal somatic variants may
initiate pathology that later spreads to broader brain regions.
We then assessed the distribution of these four somatic SNVs across cell types by
performing amplicon sequencing of DNA from neuronal (NeuN+), glial (NeuN-), diploid,
polyploid, and hypodiploid nuclei isolated by fluorescence-activated nuclei sorting (FANS) from
the regions in which the variants were originally identified (Extended Data Fig. 8a,b).
Interestingly, *TIA1* (p.H54R), *MATR3* (p.K594I), and *ALS2* (p.T787R) were enriched in
hypodiploid nuclei (Extended Data Fig. 8c), which likely represent apoptotic cells
(Supplementary Note)[49-52], suggesting a potential association with cell death. Surprisingly, these
three variants were identified in all cell fractions, but were more enriched in non-neuronal than
neuronal populations (Extended Data Fig. 8c,d). This finding could imply that neurons may
exhibit a cell-type-specific vulnerability to damaging somatic variants in ALS/FTD genes.
However, further research is needed to confirm and better understand these potential associations
and mechanisms. In contrast, the *TARDBP* (p.L248F) variant was validated only in the original
primary motor cortex sample at very low VAF (~0.5%) by both amplicon sequencing and
ddPCR, but was not detected in sorted cell fractions from an additional adjacent primary motor
cortex sample, indicating a highly focal event.
250
RNA-MosaicHunter identifies additional predicted deleterious somatic variants in bulk
RNA-seq data of sALS cases
To complement our targeted sequencing of neurodegenerative genes, which identified predicted
deleterious somatic variants in a small proportion of sALS and sFTD cases, we performed a
transcriptome-wide screen for somatic variants using bulk RNA-seq data to explore whether
previously unknown genes might cause disease in a mosaic state. We profiled predicted
deleterious somatic variants across all expressed genes in bulk RNA-seq data from 789
postmortem brain and spinal cord tissue samples of 143 sALS cases and 23 age-matched controls
generated by the New York Genome Center ALS Consortium (Supplementary Table 9; 81 sALS
and 11 control cases overlapped with our MIP cohort) using RNA-MosaicHunter[53], which
detects clonal somatic variants from bulk RNA-seq data using a Bayesian probabilistic model.
Due to limited bulk RNA-seq depth, this approach is sensitive only to somatic variants with
VAFs > ~5% and does not detect somatic variants at ultra-low levels.
Although we did not observe a significant increase in the burden of total or predicted
deleterious somatic variants in germline-free sALS cases (Supplementary Fig. 3), we identified
predicted deleterious somatic SNVs in *DYNC1H1* and *LMNA* in multiple CNS regions of two
sALS cases lacking other pathogenic or predicted deleterious germline/somatic variants (Fig. 5a
and Supplementary Table 10); both cases overlapped with our MIP cohort. Heterozygously
acting, generally *de novo*, variants in *DYNC1H1* and *LMNA* have been found in patients with
phenotypes resembling spinal muscular atrophy (SMA)[54-57], a motor neuron disease genetically
distinct but sharing some pathological overlap with ALS[58]. Both individuals presented with leg-
onset ALS and predominant spinal cord TDP-43 pathology (Fig. 5a-c). Amplicon sequencing
revealed broad CNS distribution of the *LMNA* (p.H566Y) variant (VAF 5.3-12.3%) and the
DYNC1H1 (p.R1962C) variant (VAF 0.1-5.2%), with the *DYNC1H1* variant extremely low in
the cerebellum (0.1%), thoracic spinal cord (0.8%) and lumbar spinal cord (0.8%) (Fig. 5d and
Supplementary Table 8). Notably, the *DYNC1H1* (p.R1962C) variant was undetectable in
cultured fibroblasts from the patient (Supplementary Table 8), indicating a late developmental,

CNS-restricted origin. The broad CNS distribution of these variants aligns with our previous
observations that somatic variants above ~5% VAFs are typically detected across the CNS[59],
with lower regional VAFs potentially reflecting selective neuronal loss. The *DYNC1H1*
p.R1962C variant is highly pathogenic, as it completely abolishes dynein motor function *in*
vitro[60], and causes severe malformations of cortical development and delayed psychomotor
development in patients carrying this germline variant[61,62]. Although the *LMNA* (p.H566Y)
variant was not previously reported, germline *LMNA* variants cause autosomal dominant
laminopathies with early lethality, including Hutchinson-Gilford progeria and congenital
muscular dystrophy[63,64]. Thus, germline variants in both genes would ordinarily preclude ALS,
but the mosaic state may allow for a normal early life followed by late-onset neurodegeneration.
These findings suggest that further genome-wide exploration of brain tissue for somatic variants
could reveal additional ALS genes that cause early lethality in the germline state.
290
Somatic *C9orf72* repeat expansions detected by targeted long-read sequencing
The germline *C9orf72* repeat expansion is the most common genetic cause of ALS and FTD[27,28].
Repeat lengths > 30 are generally considered pathogenic, with most affected individuals carrying
hundreds to thousands of repeats. Intermediate expansions with 20-30 repeats, confer an
increased disease risk, while expansions with < 20 repeats are regarded as normal[27,65]. While,
instability of pathogenic repeat expansions (> 30 repeats) across tissues within the same
individual has been reported[66-68], somatic expansion from non-expanded alleles has not been
described. In our ALS and FTD cohort, the repeat-primed PCR analysis identified four ALS and
FTD cases with both wild-type and expanded alleles showing reduced repeat peak heights[69]
(Extended Data Fig. 9), suggestive of somatic expansion. Targeted long-read sequencing in brain
tissues from these cases confirmed the presence of both wild-type and expanded alleles with
dozens to thousands of repeats. Remarkably, one FTD case carried two short wild-type alleles (4
and 9 repeats), together with highly expanded pathogenic alleles (Fig. 6a). Haplotype phasing
based on heterozygous single-nucleotide polymorphisms (SNPs) flanking the repeat expansions
showed that all expanded alleles shared the same SNP haplotype as the 4-repeat allele (Fig. 6b),
strongly suggesting spontaneous somatic *C9orf72* repeat expansion of a non-expanded allele,
though future analysis of more cases is needed to further characterize this possibility. In the
remaining three cases, one wild-type allele and a spectrum of expanded alleles were detected,
with minimum repeat sizes of 20, 56, and 64 repeats (Extended Data Fig. 10), consistent with
somatic expansion from intermediate or short pathogenic alleles, although the possibility of
contraction cannot be excluded. Together, these findings suggest that somatic *C9orf72* repeat
expansions may contribute to sporadic ALS and FTD. Nevertheless, even long-read sequencing
lacked sufficient sensitivity to comprehensively assess all germline-free cases (see Discussion),
and the prevalence of somatic *C9orf72* expansions remains to be determined.
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Discussion
Our data provide several important insights into sALS and sFTD. First, approximately 30% of
disease cases carried pathogenic or predicted deleterious germline variants in ALS/FTD genes,
although our pathogenicity prediction might be overestimated. This advocates for a shift from
family history-based to genetic testing-based disease classification, consistent with recently
published guidelines for ALS genetic testing and counseling[70]. Second, a small but significant
fraction (~2.1%) of germline-free sporadic cases harbored predicted deleterious somatic variants
in known ALS or FTD genes, with disease- and region-specific distributions in ALS, providing

proof of concept for a contribution of somatic variants to disease pathogenesis. In addition, we
identified genes associated with severe pediatric degenerative diseases that may contribute to
ALS in the somatic state, broadening the spectrum of potential disease genes. Finally, targeted
long-read sequencing revealed somatic *C9orf72* repeat expansions arising from wild-type,
intermediate, and short pathogenic alleles in ALS and FTD cases.
329   While the case-control enrichment of somatic variants supports a pathogenic role, these
variants occurred at surprisingly low VAFs and showed topographic restriction consistent with
focal disease onset. Such variants likely arose late in development and were not shared by other
tissue regions, exemplified by the *TARDBP* (p. L248F) variant detectable only at its site of
discovery. The focal nature of these events would preclude detection by routine genetic testing
with blood or other peripheral samples and support a mechanism by which degeneration may
spread from a site containing mutant cells to anatomically connected regions, perhaps via
intercellular transmission of TDP-43 proteinopathy[12-18,71,72]. The identification of predicted
deleterious somatic variants in the primary motor cortex or spinal cord from individuals with
ALS suggests that ALS may initiate in UMNs or LMNs before involving both.
339   While performed on a limited number of variants, our cell-type analysis revealed that
several predicted deleterious somatic variants were shared across cell types and enriched in glia
compared to neurons. However, this apparent glial enrichment may reflect preferential loss of
neurons carrying these variants, consistent with their enrichment in hypodiploid cells, a
population likely associated with apoptotic cell death. Together, these findings support a model
in which deleterious somatic variants contribute to focal disease initiation, with subsequent
neuronal loss leading to reduced VAFs over time.
346   Although only ~2.1% of germline-free disease cases carried predicted deleterious somatic
variants in ALS/FTD genes in our MIP sequencing data, this is likely greatly underestimated due
to limited sensitivity for detecting ultra-low-VAF variants (Extended Data Fig. 1). Detecting
such variants remains technically challenging[46], and broader sampling across CNS regions is also
limited. However, future duplex sequencing approaches promise the ability to define the minimal
VAFs capable of initiating disease. On the other hand, several of our statistical comparisons
approached nominal significance, highlighting the need for larger cohorts and more sensitive
approaches.
354   While MIP sequencing did not allow for detection of somatic *C9orf72* repeat expansions
in our samples, targeted long-read sequencing identified ALS and FTD cases with somatic
expansions of varying repeat sizes, including an FTD case carrying two wild-type and highly
expanded alleles, consistent with *de novo* expansion. These findings suggest that somatic
C9orf72 repeat expansions may contribute to sporadic ALS and FTD and could exhibit cell-type-
specific patterns of expansion, analogous to somatic *HTT* CAG repeat expansion in striatal
projection neurons[73]. However, the current targeted long-read sequencing approach preferentially
enriches shorter alleles and provides limited flanking sequence, precluding accurate VAF
estimation and robust haplotype phasing. Future targeted long-read sequencing with extended
flanking coverage will be required to determine the origin and prevalence of somatic *C9orf72*
repeat expansions.
365   Finally, the identification of somatic SNVs in *DYNC1H1* and *LMNA* suggests that genes
predisposing to ALS and FTD in the somatic state may extend beyond those identified in
germline studies. Although certain alleles in these genes cause motor neuron degeneration in the
form of SMA, other alleles—including *DYNC1H1* p. R1962C[61,62]—cause severe pediatric
disease that would preclude late-life ALS in the germline but may be compatible with disease in

the somatic state. Similarly, enrichment in germline-free FTD cases emerged only when all
targeted neurodegenerative genes were considered, hinting at a wider array of FTD-related
genes. Together, these findings highlight the potential of future genome-wide approaches to
uncover additional somatic genetic mechanisms and illuminate the topographic spread of
pathology from focal origins.
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Author Contributions

Z.Z., Junho K., A.Y.H., E.A.L., C.L.-T. and C.A.W. conceived and designed the study. Z.Z. performed tissue processing, MIP panel sequencing, cell sorting, amplicon sequencing, and targeted long-read sequencing. Junho K. performed bioinformatic analysis for MIP and validation data with assistance from R.D., T.S., and B.Z. A.Y.H. performed bioinformatic analysis for bulk RNA-seq data with assistance from J.P. and M.B. Jinhyeong K. performed bioinformatic analysis for long-read sequencing data. M.N. optimized and performed immunofluorescent imaging and quantification, and generated data shown in this manuscript. Z.Z., M.B.M. and R.D. designed the MIP panel. B.C., K.M., R.C.Y., and C.-Z.L. helped with tissue processing and amplicon sequencing. C.K. provided technical support for MIP sequencing. J.E.N. contributed tissue procurement and ethics expertise. T.O. and J.R. provided immunofluorescent images and interpretation of disease pathology. L.W.O. and O.A. provided

422 fresh frozen human tissues and interpretation of disease pathology. C.L.-T., E.A.L. and C.A.W.
423 supervised the study. Z.Z., Junho K., A.Y.H., C.L.-T., E.A.L. and C.A.W., wrote the manuscript.
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Competing Interests

C.L.-T serves on the scientific advisory board and/or has received consulting fees from SOLA Biosciences, Libra Therapeutics, Arbor Biotechnologies, Dewpoint Therapeutics, Mitsubishi Tanabe Pharma Holdings America, Sanofi, AUTTX LLC, Carthera, the Milken Institute, and Applied Genetic Technologies Corporation. E.A.L. serves on the scientific advisory board of Genome Insight. C.A.W. is a paid consultant (cash, no equity) to CAMP4 Therapeutics (cash, no equity), Maze Therapeutics (cash and equity), is a member of the scientific advisory board for Bioskryb Genomics, Inc., (cash) and is a founding advisor for Mosaica Medicines (equity). These companies did not fund and had no role in the conception or performance of this research project.

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Fig. 1 | Experimental and analysis strategies. **a**, Overall scheme of the experiments. Genomic
DNA isolated from 1,787 postmortem tissue samples of multiple brain regions and spinal cords
of 144 control, 291 ALS, and 117 FTD cases were used for molecular inversion probe (MIP)
capture sequencing with ultra-high depth. Created with BioRender.com. **b**,**c**, Mean sequencing
depth (**b**) and number of tissue samples (**c**) in different brain regions and spinal cords of control,
ALS, and FTD cases. Control, $n = 516$; ALS, $n = 938$; FTD, $n = 375$. Note that 42 samples from
9 ALS-FTD cases were included in both conditions. CB, cerebellum; PMC, primary motor
cortex; PFC, prefrontal cortex; PreMC, premotor cortex; SC, spinal cord; OC, occipital cortex;
AC, anterior cingulate cortex. Error bars, 95% CI. **d**, Methodological pipelines to identify
germline and somatic variants. Germline variants were called by GATK HaplotypeCaller.
C9orf72 genotype of ALS and FTD cases were determined by repeat-primed PCR. Somatic
variants were called by RePlow, MuTect2, and Pisces. Additional somatic variants were called
from 789 bulk RNA-seq profiles of multiple brain regions and spinal cords of ALS cases
generated by the New York Genome Center ALS Consortium using RNA-MosaicHunter.
Created with BioRender.com.
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Fig. 2 | *C9orf72* **repeat expansion and pathogenic germline variants in ALS/FTD genes are**
prevalent in ALS and FTD. a, Proportions of ALS and FTD cases with *C9orf72* repeat
expansion, pathogenic, and predicted deleterious germline variants in ALS/FTD genes. Cases
with multiple predicted deleterious variants are indicated with '+' sign. Note that cases with a
heterozygous variant in recessive genes (e.g., *ALS2*, *ATP13A2*) are not included in the
proportions. **b**, Distribution of *C9orf72* repeat expansion, pathogenic, and predicted deleterious
germline variants in ALS/FTD genes classified by variant types. **c**, Ranking of the top 10
mutated ALS/FTD genes. **d**, Visualization of ALS and FTD cases (vertical columns) with
pathogenic and predicted deleterious germline variants (horizontal rows) in ALS/FTD genes.
Color codes indicate the types of variants. Rectangular outline represents predicted deleterious
variants. Genes are grouped by their known involvement in the diseases. * indicates cases with
multiple pathogenic or predicted deleterious variants.
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Fig. 3 | Somatic variants in MIP sequencing data tend to be focal, protein-altering, and are
almost exclusively restricted to disease cases. a, The observed VAFs of somatic variants in
amplicon sequencing validation were consistent with the VAFs in original MIP sequencing.
Sixty-four somatic variants were validated and included in the plot. Shaded band, 95% CI of the
fitted linear regression. **b**, Total somatic variant counts classified by the number of brain regions
in which a given variant was identified. **c**, Distribution of somatic variants in all
neurodegenerative genes. Genes are categorized into four groups: ALS/FTD-related dominant
genes, ALS/FTD-related recessive genes, Tau proteinopathy-related genes, and other
neurodegeneration/dementia-associated genes. Color codes indicate variant types. Note that
somatic variants identified in controls are unlikely to alter function, with just one missense
variant (red) and the remaining being synonymous or noncoding substitutions.
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Fig. 4 | Somatic variants are enriched in ALS and FTD cases and disease-related tissue regions. **a**, Enrichment of somatic variants in different genomic regions of germline-free ALS and FTD cases compared to normal controls. **b**, Enrichment of exonic and protein-altering somatic variants in two different groups of disease-related genes (ALS genes and FTD genes) compared to normal controls. **c**, Enrichment of somatic variants in ALS/FTD genes across different brain regions of germline-free ALS and FTD cases compared to normal controls. Missing points indicate no variants were observed. **a-c**, Disease effect sizes were estimated using an individual-wise linear mixed-effects model by comparing variant burden per germline-free disease individual with that of normal individuals (see Methods). Significance and 95% CI were estimated while controlling for potential confounding factors, including average read depth, sex, post-mortem interval, sequencing batch, and number of samples per donor. Sample sizes were: control, $n = 144$; ALS, $n = 216$; FTD, $n = 78$ (biological replicates). Unadjusted P values are shown, as the tests examine biologically distinct but non-independent categories with differing background mutation structures. Created with BioRender.com. **d**, Regional distribution of VAFs of several predicted deleterious somatic variants in individual brains and spinal cords. Brain cortex is annotated by Brodmann areas. The color spectrum indicates the VAFs of somatic variants in amplicon sequencing. White indicates regions without the somatic variants. Red highlight indicates the region of initial detection by MIP sequencing. Created with BioRender.com.

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Fig. 5 | Somatic variants in *DYNC1H1* and *LMNA* in sALS. a, Two deleterious somatic SNVs
that were shared by multiple tissue regions of the ALS cases. **b**, Sections of the lumbar spinal
cord, motor cortex, and hippocampus of the two sALS cases stained with a phospho-TDP43
antibody. Scale bar, 40 μm. Arrowheads indicate the cells shown in the insets, which are
magnified to twice their original size. **c**, Quantification of phospho-TDP43 staining of CNS
tissue sections of the two sALS cases with *DYNC1H1* and *LMNA* somatic variants. For
DYNC1H1, tissue-section-level biological replicates included 3 from PFC, 4 from SC, and 5
from MC, HC, TC, PC, and OC. For *LMNA*, tissue-section-level biological replicates included 2
from SC and 5 from PFC, MC, and HC. Bar graph, mean ± 95% CI. PFC, prefrontal cortex; MC,
primary motor cortex; HC, hippocampus; TC, temporal cortex; PC, parietal cortex; OC, occipital
cortex; SC, spinal cord. **d**, Regional distribution of VAFs of somatic variants in individual brains
and spinal cords. Brain cortex is annotated by Brodmann areas. The color spectrum indicates the

VAFs of somatic variants in amplicon sequencing. White indicates regions without the somatic
variants. Created with BioRender.com.
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Fig. 6 | Somatic *C9orf72* repeat expansion in an FTD case. a, Waterfall plot showing targeted
long-read sequencing results from the prefrontal cortex tissue of a sporadic FTD case. Two wild-
type alleles with 4 and 9 repeats, respectively, are observed, along with somatic *C9orf72* repeat
expansions ranging from 748 to 2,297 repeats. Each row represents an individual long read; the
C9orf72 repeat region is shown in blue, and the flanking regions are shown in green. Red dashed
lines mark the sizes of the wild-type alleles. The *x*-axis denotes the number of GGGGCC
hexanucleotide repeats. **b**, IGV screenshot showing haplotype phasing based on SNPs flanking
the *C9orf72* repeat expansion. Long reads are grouped by repeat size category (expanded, 9
repeats, 4 repeats). SNPs rs73643418, rs73643419, and rs11789520 are heterozygous in this
individual. All expanded alleles share the same SNP haplotype as the 4-repeat allele.
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Methods

Tissue sources and sample preparation

Fresh frozen postmortem human brain and spinal cord tissues were collected by the
Massachusetts Alzheimer's Disease Research Center, Oxford Brain Bank, Target ALS
Foundation, and NIH NeuroBioBank (Supplementary Table 1) according to their respective
institutional protocols, written authorization and informed consent; they were subsequently
obtained for this study with the approval of the Boston Children's Hospital Institutional Review
Board. Research on these deidentified specimens and data was performed at Boston Children's
Hospital with approval from the Committee on Clinical Investigation. Sporadic ALS and FTD
cases were selected based on available clinical records. ALS and FTD cases without clear
recording of family histories were also selected if the age of death was above 45 years old.

gDNA of these tissue samples was extracted using the EZ1 Advanced XL (Qiagen)
system followed by an additional purification using AMPure XP beads (Beckman Coulter).

MIP panel design

A double-stranded DNA MIP panel targeting 1.4 Mb across exons and exon-intron junctions of
88 neurodegenerative genes was designed using custom scripts incorporating MIPgen[74] using the
human reference genome, hg19, with Mly1 restriction sites masked with 'N' using bedtools. The
final panel of 26,439 MIPs captures an average fragment length of 209 bp, including the
extension and ligation arms to ensure overlapping of the forward and reverse sequencing read.
The panel successfully targets 92.7% of bases including flanking intronic regions, with > 98% of
exonic bases covered with an average of at least 2 unique MIPs. All MIPs were designed to
include a custom backbone consisting of primer binding sites and dual 5-nt unique molecular
identifier (UMI). MIPs were rebalanced in the pool based on the percent of GC content within
the regions to improve coverage at GC-rich regions. We increased the copied MIPs that bind to
GC-rich regions with the following criteria: 60-70% GC = 2 copies; 70-80% GC = 5 copies; 80-
90% GC = 8 copies; > 90% GC = 10 copies. Common primer binding and Mly1 restriction
enzyme sites were added to both ends of the oligo sequences to enable blunt-end removal of the
primer binding sites. The forward and reverse compliment sequences were printed into a single
ssDNA pool by CustomArray (Bothell, WA). The resulting panel was amplified at a low cycle
number (12×), digested with Mly1 enzyme for 12 h at 37 °C, and purified using Qiagen
Nucleotide removal kit.

MIP capture and library construction

Two hundred fifty ng of gDNA was first hybridized in a 15 μl reaction with 1.5 μl of
Ampligase® 10× Reaction Buffer (VWR), 1.5 μl of the reverse blocking oligo (5'-
NNNNGAAGTCGAAGGGCTATAGGCTGCCATCACANNNN-3') and the MIP pool at 63
nM for 10 min at 95 °C and 24 h at 60 °C. Gap-fill/ligation was then performed by adding 1 unit
of Phusion™ High-Fidelity DNA Polymerase (Thermo Fisher), 4 units of Ampligase® DNA
Ligase (Epicentre), 0.2 μl of Ampligase® 10× Reaction Buffer, 0.6 μl of dNTPs (10 mM) and 1
μl of nuclease-free water to the MIP capture product and incubated at 60 °C for 1 h. For
exonuclease digestion, 50 units of Exonuclease III (Thermo Fisher), 10 units of Exonuclease I
(Thermo Fisher), 0.2 μl of Ampligase® 10× Reaction Buffer (VWR), and 2.05 μl of nuclease-
free water was added to the Gap-fill/ligation product, which was incubated for 40 min at 37 °C
and 5 min at 95 °C. Ten μl of the captured library is amplified in a 50 μl final reaction by adding
1 unit of Phusion Hot Start II DNA Polymerase (Thermo Fisher), 10 μl of 5× HF buffer, 1 μl of

dNTPs (10mM), 1 μl of the universal MIP barcode forward primer (10 μM), 1 μl of the
individual barcode reserve primer (10 μM) and 26.5 μl of nuclease-free water. MIP library
amplification was then performed under the following conditions: 98 °C for 30 s; 16 cycles of
98 °C for 10 s, 60 °C for 30 s and 72 °C for 30 s; 72 °C for 2 min. MIP library was then purified
using 2× AMPure XP Beads (Beckman Coulter,) and quantified by the Quant-iT™ dsDNA
Assay HS Kit (Thermo Fisher). Ninety-six MIP libraries were pooled together and sequenced on
one lane of Illumina Hiseq X.
757
Pre-processing and read mapping of MIP sequencing data
MIP sequencing primers were removed first from the raw FASTQ files using Cutadapt[75] (v2.4;
5' adapter of the first read: CATACGAGATCCGTAATCGGGAAGCTGAAG; 3' adapter of the
first read: ACACTACCGTCGGATCGTGCGTGT; 5' adapter of the second read:
GCTAAGGGCCTAACTGGCCGCTTCACTG; 3' adapter of the second read:
CTTCAGCTTCCCGATTACGGATCTCGTATG). Trimmed reads were aligned to the human
reference genome (GRCh37) using BWA-mem[76] (v0.7.15) and sorting and indexing were
performed using samtools[77] (v1.3.1). From the aligned BAM file, off-target reads were removed
by checking the overlaps with the target regions using bedtools[78] (v2.26.0). MIP arm regions
were masked by soft-clipping for each read using BAMClipper[79] (v1.0.1). UMI information was
extracted, and then mapped reads were deduplicated based on the mapping coordinate and the
shared UMI using UMI-tools[80] (v1.0.0). Base quality score recalibration and local realignment
were performed using the Genome Analysis Toolkit (GATK, v3.7)[81], generating final analysis-
ready BAMs.
772
Variant calling for germline variants
Initial candidates of germline SNVs and indels were identified using GATK HaplotypeCaller
with default parameter settings. Low-quality candidates were filtered out if any of the following
conditions is not satisfied: (1) ≥ 10 variant-supporting reads; (2) ≥ 20 total read-depth at the
variant site; (3) VAF ≥ 0.3; (4) GATK QUAL ≥ 50, and; (5) identified in all brain regions from
the same individual except for the samples failed to cover the variant site (<10 reads). Possible
pathogenic germline variants were further selected by satisfying all the following conditions: (1)
present in less than 0.1% of the population in any ancestry group of public databases including
dbSNP[82], the 1000 Genomes Project[83], the Exome Aggregation Consortium (ExAC)[84], the
Genome Aggregation Database (gnomAD)[85], the NHLBI Exome Sequencing Project
(ESP6500)[86], the Greater Middle East variome project (GME)[87], and Kaviar database[88]; (2)
candidates observed only in disease or control groups but not in both; (3) possible protein-
altering candidates (missense, nonsense, frame-shift, or splicing variants), and; (4) affecting 30
ALS- and FTD-related genes. Deleteriousness prediction module (see computational prediction
of variant deleteriousness section below) was then applied to the remaining candidates, and
predicted deleterious variants were reported as final deleterious germline variants. ANNOVAR[26]
was used to annotate the genomic region, affected genes, population allele frequency, and exonic
variant functions. SpliceAI[89] was additionally utilized to identify more splice-altering variants.
Candidates with delta score > 0.5 were considered to be possible splicing variants.
792
Repeat-primed PCR assay for *C9orf72* repeat expansion genotyping
Repeat-primed PCR (RP-PCR) of the *C9orf72* repeat expansion was performed in a 30 μl PCR
reaction with 150 ng of gDNA, 15 μl of 2× FastStart™ PCR Master (Roche), 2 μl of DMSO, 5 μl

of 5× Q-solution (Qiagen), 1 μl of 5 mM 7-deaza-dGTP (NEB), 1 μl of 25 mM MgCl$_2$ (Qiagen)
and 1 μl of the primer mix (40 μM of the Forward primer: 5'-/56-
FAM/AGTCGCTAGAGGCGAAAGC-3'; 20 μM of the Reverse primer: 5'-
TACGCATCCCAGTTTGAGACGGGGGGCCGGGGCCGGGGCCGGGG-3'; 40 μM of the
Anchor/tail primer: 5'-TACGCATCCCAGTTTGAGACG-3'). The reaction was performed with
touchdown PCR cycling conditions consisting of 15 min at 95 °C, followed by cycles of 94 °C
for 1 min, annealing starting at 70 °C for 1 min, and extension at 72 °C for 3 min, ending with a
final extension step of 10 min at 72 °C. The annealing temperature was decreased in 2 °C steps
as follows: 70 °C for two cycles, 68 °C for three cycles, 66 °C for four cycles, 64 °C for five
cycles, 62 °C for six cycles, 60 °C for seven cycles, 58 °C for eight cycles, and 56 °C for five
cycles. The RP-PCR products were separated by the SeqStudio Genetic Analyzer (Thermo
Fisher) with the GeneScan™ 600 LIZ™ Dye Size Standard (Thermo Fisher). Results of
fragment sizes were analyzed by Peak Scanner™ Software v1.0 (Thermo Fisher).
809
Somatic variant calling from MIP sequencing data
Three different callers RePlow (v1.1.0)[38], Mutect2 (v4.1.5)[39], and Pisces (v5.2.11)[40] were used to
generate initial candidate sets. RePlow is optimized for detecting low-VAF somatic variants
from deep targeted sequencing data. It generates a profile of background errors per substitution
type and utilizes these distributions as priors in a Bayesian model to estimate the probability of a
variant candidate being an error. MuTect2 is one of the most widely used somatic variant callers,
particularly sensitive to detecting low-VAF variants in impure and heterogeneous samples. It
employs a Bayesian classifier to evaluate the likelihood of a variant being genuine versus a
sequencing error and applies multiple filters to reduce false positives. Pisces is specifically
designed for detecting somatic mutations from amplicon sequencing data, particularly in cases
where no matched control sample is available. It stitches paired-end reads into consensus reads
and recalibrates variant quality scores specifically tuned to address amplification-related errors
such as thermal damage or deamination, often observed in FFPE samples.
Each sample was analyzed by all three callers using single-sample mode. Default
parameter settings were used except for the adjustments for disabling the coverage limit.
Variants that passed all the filters from each caller were used to make three different initial sets.
Candidates identified by only one caller were discarded, and those called at least two callers
were retained as a double-call set. For indels, double-calls between Mutect2 and Pisces were
used as somatic indel candidates since RePlow does not support indel detection. For SNVs,
among double-calls Mutect2-Pisces pairs were additionally filtered out due to high false positive
rates and low validation rates in the benchmarking data set (Extended Data Fig. 3d). Remaining
RePlow-based SNV double-calls and indel candidates were subject to multi-step variant filters to
further remove false positive candidates.
Unlike germline variant calling, somatic variant calling aims to reliably detect low-VAF
variants up to ~0.5%, which requires enough supporting evidence to control the false positive
rate. Calling thresholds such as variant-supporting read count, read-depth at the variant site, and
average base-call quality were determined based on the benchmarking data. Somatic variants
were selected satisfying all the following conditions: (1) ≥ 50 total read-depth at the variant site;
(2) ≥ 15 variant-supporting reads excluding the reads with the variant allele on their probe-arm
regions; (3) > 30 average base-call quality of variant allele; (4) ≥ 2 different types of variant-
supporting amplicons; (5) 0.001 ≤ VAF ≤ 0.4; (6) ≤ 3 variant candidates within 20-bp window

from the same sample; (7) present in less than 0.1% of the population in any ancestry group of
public databases, and; (8) observed in no more than two different individuals.
Somatic variants were further annotated with similar criteria for selecting predicted
deleterious germline variants. Among the final candidates, variants that are (1) observed only in
disease or control groups but not in both, (2) possible protein-altering variants, and (3) affecting
ALS- and FTD-related genes were selected and applied for the deleteriousness prediction
module. ANNOVAR and SpliceAI were utilized to annotate variants with various genomic
information and detect additional splice-altering variants, respectively.
849
Computational prediction of variant deleteriousness
Deleteriousness prediction module was applied to filtered germline and somatic variants to refine
the predicted deleterious (potentially pathogenic) candidate sets. Variants that were previously
reported as benign/likely benign in the clinical databases (ClinVar[90] and Human Gene Mutation
Database[91]) were excluded from the predicted candidate set. Nonsense, frameshift, and
canonical splicing variants (±1-2 splice sites) were assumed to disrupt gene function and were
included in the predicted deleterious set. For missense variants, the dbNSFP database[92] was
utilized to adopt multiple computational algorithms (SIFT[93], PolyPhen2[94], LRT[95],
MutationTaster[96], MutationAssessor[97], FATHMM[98], FATHMM-MKL[99], PROVEAN[100],
MetaSVM[101], MetaLR[101]), considering damaging effects at different levels such as biochemical
property, protein structure, and evolutionary conservation. Categorical prediction results of each
algorithm were delivered by ANNOVAR. A missense variant was selected to be predicted
deleterious if at least three different algorithms predicted damaging effects (deleterious for SIFT,
LRT, FATHMM, PROVEAN, MetaSVM and MetaLR; probably damaging for PolyPhen2;
disease-causing for MutationTaster). Possibly/likely damaging predictions were excluded for
more conservative selection. For ALS/FTD-related genes, previously reported inheritance
patterns (dominant/recessive) were carefully checked. For recessive genes, two independent
variants in the same gene were required to determine whether a given individual was affected by
predicted deleterious variants.
869
Benchmarking with spike-in datasets
Extracted gDNA from two Coriell cell lines (GM12878 and GM24695) were used to generate a
spike-in data. Extracted DNA were mixed at five different levels to mimic low-level somatic
variants, targeting the VAFs of 0.5%, 1%, 2.5%, 5%, and 10%. Genomic DNA from GM12878
cells was spiked into DNA from GM24695, therefore unique germline SNPs in GM12878 were
served as somatic variants. Genomic position and genotype information for germline SNPs of
Coriell samples were obtained from NIST high-confidence call sets[102]. A total of 165 SNPs (57
homozygous and 108 heterozygous SNPs) were initially covered by our designed MIP panel.
Among them, we only utilized heterozygous SNPs for benchmarking to accurately reflect target
VAFs. Additionally, we excluded the spike-in variants from benchmarking if (1) a given site
could be covered by only one amplicon, or (2) have read-depth of less than 100, as these are low-
quality variants unable to be detected due to inherent issues with panel design rather than the
variant calling process. RePlow, Mutect2, Pisces, and their combinations were tested. Detected
variants not in the benchmark set were considered to be false positives, except for GM24695
germline SNPs and excluded spike-in variants.
885
Somatic variant calling from bulk RNA-seq data

Raw bam files of bulk RNA-seq and matched WGS data for sALS and control cases of the New
York Genome Center ALS Consortium were obtained from the New York Genome Center.
RNA-seq reads extracted from raw bam files were aligned to the GRCh38 human reference
genome by STAR (v2.5.0a)[103] in the two-pass mode with the reference gene annotation
(Gencode version 39). The aligned bam files were processed by Picard (v1.138) to remove
duplicates, and then by GATK (v3.6)[104] for SplitNCigarReads, indel realignment, and base
quality recalibration. We further excluded reads that were improperly paired or with ambiguous
alignment.
Somatic SNVs were called by RNA-MosaicHunter (v1.0) with default parameters.
Derived from MosaicHunter[105], which was designed for somatic variant calling in DNA
sequencing, RNA-MosaicHunter incorporates a Bayesian genotyper and a series of empirical
filters to systematically distinguish somatic variants from technical artifacts and germline
variants, with 59% sensitivity and 94% precision benchmarked using cancer datasets.
Specifically, germline variants identified from the matched WGS data from the same individual
were excluded. We excluded A-to-G candidates because they are most likely led by the
widespread A-to-I(G) RNA editing events in the human genome. To remove recurrent artifacts,
we only considered exonic candidates that were called in one or two individuals. We further
excluded candidates present in human polymorphism databases including dbSNP[82], the 1000
Genomes Project[106], the Exome Sequencing Project[107], and the Exome Aggregation
Consortium[108].
907
Long-read targeted sequencing
The PureTarget kit (PacBio) was used to prepare sequencing libraries with 2 μg of high-
molecular-weight DNA following the manufacturer's protocol. Pooled libraries were sequenced
on one Revio SMRT Cell. Raw HiFi reads were aligned to the human reference genome
(GRCh38) using minimap2 (v2.28)[109], and sorting and indexing were performed using samtools
(v1.3.1)[77]. The Tandem Repeat Genotyping Tool (TRGT, v1.1.1)[110] was used to determine the
C9orf72 repeat counts for each sample (chr9:27573528-27573546, (GGGGCC)n). A waterfall
plot was generated using TRVZ to visualize repeat counts with mosaicism.
916
Nuclei isolation and whole genome amplification
Isolation of total (DAPI+), neuronal (NeuN+), non-neuronal (NeuN-), and damaged (low DAPI)
nuclei were achieved by FANS together with nuclear staining of NeuN (Millipore, MAB377,
clone A60, 1:1,500) and DAPI following a previously published study[111]. Five hundred nuclei of
each cell population were sorted into wells of 96-well plates.
Sorted nuclei were subjected to genome amplification using the Primary Template-
directed Amplification kit (BioSkryb, 100136) following the manufacturer's protocol.
924
Amplicon sequencing
Primer sets targeting each identified somatic SNV were designed using BatchPrimer3
(Supplementary Table 11). Amplicon was amplified for 25 cycles in a 50 μl PCR reaction with
50 ng of gDNA, 1 unit of Phusion Hot Start II DNA Polymerase (Thermo Fisher), 10 μl of 5×
HF buffer, 1 μl of dNTPs (10 mM) and 10 μl of each primer (10 μM). Amplicon PCR products
were then purified by a 0.65× + 1.05× double size selection with AMPure XP Beads (Beckman
Coulter, A63882). Purified amplicons were then pooled based on the concentrations measured by

the Quant-iT™ dsDNA Assay HS Kit (Thermo Fisher) and sequenced using Amplicon-EZ
(Genewiz).
934

Immunohistochemistry
Immunohistochemistry was performed using DAB (3,3'-Diaminobenzidine) detection following
published protocols[112]. Briefly, 7-µm formalin-fixed, paraffin-embedded (FFPE) sections were
dewaxed using citrisolve, before being rehydrated through decreasing concentrations of ethanol.
Antigen retrieval was performed using sodium citrate buffer pH 6.0 at 121 °C for 15 min.
Endogenous peroxidases were blocked using 3% hydrogen peroxide solution, and non-specific
binding was blocked using 10% normal goat serum. Sections were then incubated overnight at
4 °C with primary antibody (pTDP-43 mouse polyclonal, CosmoBio CAC-TIP-PTD-P03,
1:10,000). After washing with TBS-Triton, sections were incubated with a Horseradish
peroxidase (HRP)-conjugated Goat anti-mouse secondary (Dako) for 1 h at room temperature.
HRP signal was detected using DAB substrate (Dako) applied for 15 min. Counterstaining was
performed using Coles hematoxylin for 1 min. Sections were then dehydrated, cleared using
citrisolve, and mounted using glass coverslips. All sections were viewed using a Leica upright
light microscope and assessed for section quality prior to whole-slide digital scanning.
949

Quantification of p-TDP43 burden by immunohistochemistry
Stained sections were scanned using a NanoZoomer whole-slide digital imager at 40X
magnification. Images were then visualized and quantified using QuPath image analysis software
and algorithms following published protocols[112]. Briefly, for cortical/cerebellar sections 5 ROI
measuring 3 mm^2 (1000 × 3000 µm) were placed equidistantly around a single gyrus with the
short end of the ROI placed at the pial surface. Pathology was then quantified using a positive
pixel count within each ROI and measurements were averaged to provide an output of positive
pixels/mm^2. For spinal cord sections, square ROI (2.25 mm^2) was placed on each side of the
central canal within the anterior horn and measurements were averaged.
959

Data Availability
The bulk RNA-seq data generated by the NYGC ALS Consortium are available through
controlled access via the Target ALS Data Portal (https://dataengine.targetals.org/). Access
requires acceptance of the Target ALS Data Use Agreement and submission of a data access
request through the portal application process. Additional details are provided in the Target ALS
Data Portal User Manual. The MIP-based targeted sequencing data and long-read sequencing
data generated in this study have been deposited in dbGaP under accession number phs003530,
with access governed by human subject privacy regulations. Germline and somatic variants
identified and validated in this study are listed in the supplementary tables.
969

970 **Code Availability**
971 The source code and default configuration file of RNA-MosaicHunter have been published and
972 are available at https://github.com/AugustHuang/RNA-MosaicHunter. Other custom analysis
973 scripts used in this study are publicly available at GitHub (https://github.com/kimjh607/ALS-
974 FTD-somatic-mosaicism) and archived at Zenodo (https://doi.org/10.5281/zenodo.18682277).
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