Translation of upstream open reading frames does not trigger nonsense-mediated decay

in Arabidopsis (99/120 characters)

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ABSTRACT (199/200 Words)

Upstream ORFs (uORFs) are widespread cis-regulatory elements in the 5' untranslated regions

of eukaryotic genes. In addition to repressing main ORF translation, uORF translation in animals

also reduces mRNA stability through nonsense-mediated decay (NMD). In contrast, the roles of

uORFs in plants are less understood. One limitation of most prior studies is the inclusion of both

translated and untranslated uORFs for analysis. Here, we identified actively translated uORFs

(TuORFs) in Arabidopsis through ribosome profiling and examined their roles in gene

expression by integrating multiple genome-wide datasets. Similar to animal systems, TuORFs

are associated with 38%, 14%, and 43% reductions in translation efficiency, mRNA stability, and

protein levels, respectively. Surprisingly, TuORF-containing mRNAs in plants have 51% higher

transcript levels, which is opposite to the observations in animals. We present multiple lines of

evidence showing that uORF translation generally does not trigger NMD in Arabidopsis. First,

TuORF-containing mRNAs are not upregulated in various NMD mutants. Second, TuORF-

containing mRNAs and known NMD targets have distinct expression patterns, and they are

translationally repressed via different mechanisms. Finally, TuORF-containing mRNAs and

NMD targets are degraded through separate pathways. Together, our results highlight

fundamental differences in gene regulation mediated by TuORFs in plants and animals.

INTRODUCTION

Upstream open reading frames (uORFs) are cis-acting translational repressors that regulate many eukaryotic protein-coding transcripts. Mutations or modifications of uORFs are associated with human diseases and plant traits (Calvo et al., 2009; Xu et al., 2017; Zhang et al., 2018; Lee et al., 2021). Considering AUG as a translation start codon, 30%–70% of genes contain at least one potential uORF in moss, Arabidopsis, fly, mouse, and human (Kim et al., 2007; Johnstone et al., 2016; Zhang et al., 2021). The genes with predicted uORFs are enriched for signal transduction proteins, transcription factors, and membrane proteins in animals (Zhang et al., 2021) and transcription factors and protein kinases in plants (Kim et al., 2007), underlining the potential importance of uORFs in modulating the expression of vital cellular regulators.

Many studies in eukaryotes have reported that uORF translation reduces main ORF (mORF) translation (Chew et al., 2016; Johnstone et al., 2016; Liu et al., 2013; Wu et al., 2019; Lin et al., 2019), and it is estimated to reduce protein production by 30%–80% (Calvo et al., 2009; Chew et al., 2016). According to the scanning model of mRNA translation, the 43S pre-initiation complex, which contains the small ribosomal subunit, several translation initiation factors, and initiator tRNA, binds the 5' cap and scans the 5' untranslated region (UTR) to search for an optimal initiation sequence (Hinnebusch et al., 2016). When a uORF is translated, the ribosome may stall during elongation or at termination, or it may dissociate from the mRNA after termination, thus reducing the translation of the downstream mORF (Dever et al., 2020). The mORFs downstream of uORFs may get translated through leaky scanning or reinitiation (Kozak, 1999).

uORF translation is also expected to trigger mRNA degradation, presumably through nonsense-mediated decay (NMD). NMD is an evolutionarily conserved regulatory mechanism to eliminate aberrant mRNAs, such as those that contain premature termination codons and unusually long 3' UTRs (He and Jacobson, 2015; Kurosaki et al., 2019). Two pathways have

been established in mammals: the Exon Junction Complex (EJC)-dependent and long 3' UTR

pathways. Normally, EJC and two NMD factors (UP-FRAMESHIFT SUPPRESSOR 2 and 3;

UPF2 and UPF3) deposited during splicing of mRNAs will be evicted by the ribosome in the first

round of translation. In the presence of a premature termination codon, EJC-UPF2-UPF3

complexes downstream of the termination codon remain attached to the mRNA. This allows

UPF1, which is part of the Surveillance Complex formed at translation termination, to interact

with UPF2 and UPF3 in proximity. Interaction of UPF1, UPF2, and UPF3 activates the

SUPPRESSOR WITH MORPHOLOGICAL EFFECT ON GENITALIA1 (SMG1) kinase, which

phosphorylates UPF1. This establishes the first phase of NMD, substrate recognition, and

activates downstream signaling and eventually the second phase of NMD, substrate

degradation, which involves SMG5, SMG6, and SMG7. Phosphorylated UPF1 also inhibits

translation initiation (Isken et al., 2008). Alternatively, unusually long 3' UTRs, which result in

poor translation termination, are believed to promote NMD through phosphorylated UPF1

independent of EJC (Kurosaki et al., 2014).

The plant NMD pathway is best characterized in *Arabidopsis thaliana*. Arabidopsis

possesses homologous NMD factors involved in the first phase, including UPF1, UPF2, and

UPF3, and the second phase, SMG7 (Raxwal and Riha, 2016). However, Arabidopsis is

missing the homologs of animal SMG1, SMG5, and SMG6 (Lloyd and Davies, 2013). The

absence of key NMD components in plants suggests significant variation in the NMD

mechanisms between plants and animals. It remains an open question how the differences in

NMD components affect substrate recognition and degradation in plants.

In Arabidopsis, putative NMD targets have been identified based on transcripts that

overly accumulate in NMD-defective mutants (Kurihara et al., 2009; Kalyna et al., 2012;

Drechsel et al., 2013; Degtiar et al., 2015). However, there is a concern that upregulation of

some transcripts in NMD mutants might be caused by secondary effects of NMD (Raxwal et al.,

2020). For example, NMD mutants affect numerous aspects of plant physiology, and some

alleles cause severe growth retardation and lethality due to constitutive immune responses

(Jeong et al., 2011; Rayson et al., 2012; Riehs-Kearnan et al., 2012; Filichkin et al., 2015;

Merchante et al., 2015; Chiam et al., 2019). Moreover, UPF1 has been reported to have multiple

functions in plants, such as alternative splicing regulation and translational control, in addition to

NMD (Yoine et al., 2006; Raxwal et al., 2020). To identify *bona fide* NMD targets, *upf1* and

smg7, which are defective in the first and second phases of NMD, respectively, were introduced

into *pad4*, in which the immune response is inactivated (Riehs-Kearnan et al., 2012; Raxwal et

al., 2020). The 333 transcripts upregulated in both *upf1 pad4* and *smg7 pad4* were defined as

high-confidence NMD targets. It was shown these NMD targets are translationally regulated by

UPF1 and that their turnover requires decapping and 5' to 3' decay (Raxwal et al., 2020).

106   As uORF translation could result in a transcript with an unusually long 3' UTR, uORF-

containing mRNAs are predicted to be targeted by NMD (Kertész et al., 2006; Kurosaki et al.,

2019). Consistent with this expectation, genome-wide studies in zebrafish, mouse, and human

have found that transcripts with translated uORFs have lower mRNA levels globally (Chew et

al., 2016; Johnstone et al., 2016). Moreover, transcripts containing translated uORFs are

upregulated in knockdown of NMD factors in mouse embryonic stem cells and human cell lines

(Hurt et al., 2013; Baird et al., 2018), supporting the idea that these transcripts are NMD targets.

113   In plants, whether uORF translation triggers NMD is still under debate. Two genome-

wide studies suggest that uORF-containing mRNAs are targets of NMD, as transcripts

possessing predicted uORFs are enriched in the genes upregulated in Arabidopsis *upf1* and

upf3 mutants (Kurihara et al., 2009; Drechsel et al., 2013). In contrast, three studies suggested

that NMD only controls a small fraction of uORF-containing mRNAs in plants, such as those

containing uORFs with > 35 codons (~2% of uORF-containing genes), or those responsive to

certain environmental stimuli, such as polyamine or blue light (Nyikó et al., 2009; Uchiyama-

Kadokura et al., 2014; Kurihara et al., 2018). Moreover, evidence from several gene-specific or

genome-wide studies suggests that uORFs do not trigger NMD. First, the mRNA levels of

several uORF-containing transcripts, including *NIP5;1* and *AtMHX*, are not increased in NMD

mutants (Saul et al., 2009; Tanaka et al., 2016). Second, a recent genome-wide study reported

that genes with increasing numbers of uORFs did not show lower transcript abundances in

Arabidopsis and tomato (Li and Liu, 2020). It is unknown whether some uORF-containing

mRNAs escape NMD or whether uORF-containing mRNAs are generally not targets of NMD in

plants. It has been concluded that the mRNA features studied so far is insufficient to infer NMD

targetability in plants (Ohtani and Wachter, 2019). Overall, it remains unclear how translated

uORFs affect mRNA stability and what the role of NMD in this process is in plants.

One limitation of the previous studies is that predicted uORFs, which include both

translated and untranslated uORFs, were commonly included in the analyses. Although many

uORFs have been predicted based on nucleotide sequences (Zhang et al., 2021), determining

which uORFs are actively translated still requires experimental evidence. Recently, deep

sequencing of ribosome footprints – known as ribosome profiling or Ribo-seq – was exploited to

study genome-wide mRNA translation (Ingolia et al., 2009; Brar and Weissman, 2015). Several

Ribo-seq studies in Arabidopsis have deployed different strategies to identify translated uORFs

(TuORFs) (Liu et al., 2013; Juntawong et al., 2014; Hsu et al., 2016; Hu et al., 2016; Bazin et

al., 2017; Kurihara et al., 2020). Among different parameters used, 3-nucleotide (3-nt)

periodicity, which corresponds to translating ribosomes deciphering 3 nts per codon, is

considered to be the most stringent criterion to identify high-confidence translation events

(Ingolia et al., 2009; Calviello et al., 2016; Xu et al., 2018; Harnett et al., 2021). Our previous

work and a recent study using RiboTaper (Calviello et al., 2016), which computes the statistical

significance of 3-nt periodicity, uncovered 187–1378 AUG-initiated TuORFs in Arabidopsis (Hsu

et al., 2016; Kurihara et al., 2020).

Here we further improved the Ribo-seq methodology and identified 2093 AUG-initiated

TuORFs in Arabidopsis seedlings. We systematically examined the assumed roles of TuORFs

in mORF translation and mRNA decay by integrating multiple genome-wide datasets. As

predicted, we found that TuORF mRNAs are associated with lower translation efficiency, shorter

mRNA half-lives, and lower protein abundance. Surprisingly, we found that, in contrast to the

animal paradigm, plant TuORF mRNAs have higher steady-state mRNA levels, and they are not

regulated by the NMD pathway. Examination of numerous parameters revealed that uORFs and

NMD regulate gene expression through separate mechanisms. Taken together, our results

reveal unexpected features and regulation of uORFs in Arabidopsis and suggest that plants and

animals have evolved different uORF-dependent mechanisms for gene regulation.

RESULTS

Translated uORFs identified via an improved Ribo-seq method

Previously we optimized Ribo-seq to reach high 3-nt periodicity, and we exploited this

unique feature of mRNA translation to identify translated ORFs in Arabidopsis (Hsu et al., 2016).

Although the data revealed many unannotated translation events, the number of TuORFs

uncovered was low, presumably due to insufficient Ribo-seq coverage within uORFs. We

proposed that increasing Ribo-seq coverage could improve the efficiency of translated ORF

identification, especially for relatively short ORFs, such as uORFs. Here we optimized the

coverage of Ribo-seq in 7-day-old Arabidopsis seedlings by minimizing the potential loss of

ribosome footprints during the purification process (see MATERIALS AND METHODS for

details).

In addition to high correlations among biological replicates (**Figure S1A**), our new Ribo-

seq data displayed characteristics expected for high-quality data, including strong 3-nt

periodicity (92% in-frame reads), high enrichment for coding sequences (CDSs), and

characteristic ribosome footprint lengths (**Figure 1A-C**). The improved coverage is evident when

examining the number of unique ribosome footprints detected globally. At a sequencing depth of

20 million reads from randomly selected reads, the number of unique ribosome footprints in our

new data was 4.01-fold higher than that in our previous dataset (**Figure 1D**). The coverage

improvement is also evident in individual transcript profiles within both uORFs and mORFs. For

example, a TuORF in *REPRESSOR OF GA1* (*RGA1*) was identified in our current data but

missed in our previous study despite the mRNA levels being similar (**Figure S1B**).

177   The high-coverage data increased the number of TuORFs uncovered. In total, 2093

TuORFs were identified in 1768 genes based on strong 3-nt periodicity (**Figure 1E-G** and **Table**

S1A). This result is in strong contrast to the relatively few TuORFs identified in Arabidopsis

seedling shoots (93 TuORFs) and roots (138 TuORFs) from our previous data obtained using

plants grown under similar conditions and using the same statistical analysis and annotations

(**Figure 1E** and **Table S1B-C**). The new data also revealed translation from a higher number of

Conserved Peptide uORFs (CPuORFs) and annotated coding sequences (mORFs of protein-

coding genes) (**Figure 1E** and **Table S1D-F**). Taken together, these results showed that our

enhanced-coverage Ribo-seq data dramatically improved the identification of TuORFs.

186   Previous analyses of predicted uORFs found that transcription factors and protein

kinases are overrepresented among uORF-containing genes (Von Arnim et al., 2014; Kim et al.,

2007). Consistent with these analyses, functions related to transcription factors and protein

kinases/phosphatases were identified as being enriched in our Gene Ontology (GO) analysis of

TuORF-containing genes (**Table S2**). We found that many TuORF-containing genes encode

extensively studied regulators of plant growth and environmental responses. For example, this

list included well-known transcription factors, such as those involved in hormone signaling

(RGA1 and AUXIN RESPONSE FACTOR2/3/8), root development (SCARECROW), the

circadian clock (TIMING OF CAB EXPRESSION1), and light signaling (PHYTOCHROME

INTERACTING FACTOR3/4/7) (Silverstone et al., 1998; Strayer et al., 2000; Di Laurenzio et al.,

1996; Ellis et al., 2005; Nagpal et al., 2005; Ni et al., 1998; Huq and Quail, 2002; Leivar et al.,

2008). The list also included well-characterized protein kinases, such as Phytochromes B and

E, ribosomal-protein S6 kinases (S6K1/2), and SNF1-related protein kinase 2 (SNRK2.1/2.5)

(**Figure S1B, S2A-N**) (Sharrock and Quail, 1989; Devlin et al., 1998; Zhang et al., 1994;

Mizoguchi et al., 1995; Park et al., 1993; Boudsocq et al., 2004). These results support the

notion that TuORFs control important cellular regulators in plants.

TuORF-containing genes have higher mRNA abundance but are associated with lower

TE, shorter mRNA half-lives, and lower protein abundance

To understand how uORF translation affects gene expression, we compared mRNA

abundance and translation efficiency of TuORF-containing genes (hereafter, TuORF genes),

genes containing sequence-predicted uORFs but underected by RiboTaper (hereafter, UuORF

genes), and genes containing an annotated 5' UTR but no potential uORFs (hereafter, no-uORF

genes). Note that some UuORFs may be translated but missed in our analysis because their 3-

nt periodicity is insufficient to pass the statistical test: these include extremely short uORFs,

those translated at a low frequency, and those overlapping with other uORFs, thus disrupting

the 3-nt periodicity. To study the global effects of uORFs on mRNA stability and protein

abundance, we compared genes with TuORFs, UuORFs, or no-uORFs, using available mRNA

half-life data determined by metabolic labeling and our previous quantitative proteomics data,

both also from Arabidopsis seedlings (Szabo et al., 2020; Song et al., 2018) (**Figure 2A-D**). The

growth conditions and source of the data are summarized in **Table S3**.

Studies in vertebrates have shown that uORF translation is associated with lower

steady-state mRNA abundance, consistent with the expectation that TuORF mRNAs are

regulated by NMD (Chew et al., 2016; Johnstone et al., 2016). Surprisingly, we found that

TuORF genes in Arabidopsis have higher steady-state mRNA levels than that of no-uORFs

genes (**Figure 2A**), despite having shorter mRNA half-lives as expected (**Figure 2C**). The

median mRNA level of TuORF genes was 51% higher than that of no-uORF genes. The higher

steady-state mRNA levels and shorter mRNA half-lives suggest that TuORF genes are highly

transcribed. To examine whether the high transcript levels of TuORF genes are conserved in

other plant species, we reanalyzed our previous tomato Ribo-seq data from seedling roots (Wu

et al., 2019) and found the TuORF genes in tomato also have higher steady-state mRNA levels

(**Figure 2E**). Given that Arabidopsis and tomato diverged more than 100 million years ago

(Clarke et al., 2011), our results suggest that TuORF-associated higher mRNA abundance

could be prevalent in plants. To investigate whether our observation is specific to seedlings or

common to different developmental stages/tissues/genetic backgrounds in Arabidopsis and

tomato, we examined the mRNA levels of the uORF-containing genes in 128 independent RNA-

seq studies collected by the EMBO-EBI expression atlas (Papatheodorou et al., 2020).

Consistent with our Arabidopsis seedling and Tomato seedling root data, we found the TuORF

genes are expressed at higher levels in diverse tissues, developmental stages, and ecotypes in

Arabidopsis and tomato (**Figure 2G and S3**), supporting the hypothesis that TuORF genes are

preferentially expressed at high levels in these plants under a wide range of sampling

conditions.

One trivial explanation for the higher mRNA abundance of TuORF genes is that TuORFs

are more likely to be discovered in highly abundant transcripts. To test this possibility, we first

examined genes containing CPuORFs that were identified independently from our approach by

evolutionary conservation (Hayden and Jorgensen, 2007; Jorgensen and Dorantes-Acosta,

2012; Takahashi et al., 2012; Vaughn et al., 2012; Van Der Horst et al., 2019). We found that

the median mRNA abundance of CPuORF genes was 2.1-fold higher than that of no-uORF

genes (**Figure S4**), consistent with our observations that genes with TuORFs are expressed at

higher levels. Next, we examined the distributions of TuORF and UuORF mRNA levels (**Figure**

2H). We found many UuORF genes were highly expressed as well. In fact, the distributions of

TuORF and UuORF mRNA levels in seedlings completely overlapped (**Figure 2H**), indicating

that mRNA abundance is not a major determinant of whether a uORF is detected as translated.

Finally, the fact that TuORFs are associated with lower TE, shorter mRNA half-lives, and less

protein accumulation than UuORF genes (**Figure 2B-D, F**), indicates that TuORFs are potent

inhibitors for gene expression. Together, our results support the hypothesis that TuORFs are

selectively translated to repress the translation and mRNA stability. Our observation of the high

mRNA abundance of plant TuORF genes also reveals that plants and animals implement

TuORFs differently in gene regulation.

Interestingly, even though globally TuORF genes in Arabidopsis start with 51% more

mRNAs than no-uORF genes, they produce 41%–43% less protein, on average (**Figure 2A, 2D,**

and S5), likely due to the combined effects of a 38% reduction in translation efficiency and 14%

reduction in mRNA half-life (**Figure 2B, 2C**). These observations suggest that global repression

of TuORF gene protein synthesis is determined by multilayer post-transcriptional regulation.

Notably, for the four parameters examined here, UuORF genes consistently have the levels

between those of TuORF genes and no-uORF genes: 4% increase in mRNA levels and 11%,

6%, and 21% reductions in TE, mRNA half-life, and protein abundance, respectively (**Figure**

2A-D). These results indicate that UuORFs have smaller effects on gene expression than

TuORFs and highlight the importance of distinguishing TuORFs and UuORFs in downstream

analysis.

TuORF genes are larger and likely evolutionarily older than other genes

We next investigated whether there are general structural differences among TuORF,

UuORF, and no-uORF mRNAs. Considering the directionality of ribosome scanning, we

hypothesized TuORFs might be associated with longer 5' UTRs. Indeed, we found that the

median 5' UTR of TuORF mRNAs is 2.5-fold of that of no-uORF mRNAs (**Figure 3A**), which is

consistent with a previous report that uORFs in longer 5' UTRs are more likely to be translated

(Liu et al., 2013). However, unexpectedly, we found that TuORF mRNAs also have longer CDSs

and 3'UTRs, as well as longer transcript lengths, than UuORF mRNAs (**Figure 3B-D**). No-uORF

mRNAs were the shortest in all categories examined (**Figure 3A-D**). These results support the

hypothesis that TuORFs are more likely to be found in longer 5' UTRs and that TuORF mRNAs

tend to encode larger proteins (**Figure 3B**). The enrichment of TuORF genes for transcription

factors, protein kinases, and related functions in GO term analysis (**Table S2**) also implies

TuORF genes tend to encode proteins with conserved domains. It is known that evolutionarily

older genes are generally longer, expressed at higher levels, and have protein sequences that

are under stronger purifying selection (Wolfa et al., 2009; Cui et al., 2015). Our observations

that TuORFs have higher expression levels, longer transcript lengths, and are enriched for

transcription factors and protein kinases imply that older genes are more likely to be regulated

through mRNA translation and degradation mediated by TuORFs.

TuORF mRNAs are not upregulated in the NMD mutants

The established paradigm from vertebrates is that uORF translation triggers NMD (Chew

et al., 2016; Johnstone et al., 2016; Hurt et al., 2013). Despite having higher steady state mRNA

levels (**Figure 2A**), TuORF genes in Arabidopsis also have shorter mRNA half-lives (**Figure

2C**). Therefore, we investigated whether TuORF-containing transcripts in Arabidopsis are

subject to NMD.

If TuORFs trigger NMD, one would expect TuORF mRNAs to be among the known NMD

targets and their mRNA levels to be increased in NMD-deficient mutants. We first examined

whether TuORF mRNAs were reported in a high-confidence NMD target list (Raxwal et al.,

2020). Among the 333 NMD targets that were upregulated in both *upf1 pad4* and *smg7 pad4*

adult leaves (Raxwal et al., 2020), 248 genes with an annotated 5' UTR were expressed in our

data (**Table S4A**). However, only 49 of 1748 TuORF genes were identified in the high-

confidence NMD target list (**Table S4B**). The 49 genes represent 2.8% of TuORF-containing

genes and 20% of the NMD targets. These results suggest that TuORFs generally do not trigger

NMD; rather, a notable fraction (20%) of NMD targets are regulated by both TuORFs and NMD.

One caveat is that the poor overlap between the two gene lists could be due to differences in

growth conditions, developmental stages, and sample handling in different labs (**Table S3**).

To further investigate if TuORF mRNAs are NMD targets, we examined whether they are

upregulated in NMD-deficient mutants by analyzing four published RNA-seq datasets for NMD

mutants (Drechsel et al., 2013; Merchante et al., 2015; Raxwal et al., 2020; Gloggnitzer et al.,

2014). Although these datasets were collected from different NMD mutants and at various

developmental stages (**Table S3**), the high-confidence NMD targets were consistently and

significantly upregulated in all the NMD mutants examined compared with the control plants

(i.e., 3.05-, 1.54-, 4.98-, and 2.65-fold upregulation for NMD targets without TuORFs; 3.04-,

2.23-, 5.04-, 2.66-fold upregulation for NMD targets with TuORFs) (**Figure 4A-D**). In contrast,

the mRNA levels of TuORF mRNAs showed no or little change in the NMD mutants (i.e., 1.01-,

0.98-, 1.10- and 1.10-fold difference). Thus, the NMD targets, but not TuORF mRNAs, are

dramatically increased in the NMD mutants, further supporting the idea that uORF translation in

Arabidopsis generally does not trigger NMD.

Lengths of the TuORF and 3' UTR alone are insufficient to predict NMD targetability

A previous study suggested that uORFs that encode a peptide longer than 35 amino

acids are likely to trigger NMD in Arabidopsis (Nyikó et al., 2009). To test whether TuORF

length is associated with NMD targetability, we analyzed TuORF lengths in the NMD targets and

non-NMD targets. We found that the NMD target TuORF genes tend to have longer uORF

peptides than non-NMD target TuORF genes (median length 28 amino acids vs. 20.5 amino

acids, p = 9.874e-07, Wilcoxon signed-rank test). However, the distributions of the uORF

peptide lengths in NMD targets and non-NMD targets largely overlapped (**Figure S6**), indicating

that TuORF length alone is insufficient to predict whether a TuORF triggers NMD of a transcript.

We also examined whether the high-confidence NMD targets have longer 3' UTRs, a

feature that triggers NMD. Interestingly, the NMD targets do not have significantly longer 3'

UTRs, no matter whether they contain TuORFs or not (**Figure S7**). Taken together, our results

are consistent with a recent suggestion that no definite RNA features studied to date are

sufficient to predict NMD targets in plants (Ohtani and Wachter, 2019).

TuORF mRNAs and NMD targets have distinct expression patterns

To further characterize the roles of NMD and TuORFs in gene expression regulation, we

compared the expression patterns of the NMD targets and TuORF mRNAs (**Figure 5**).

Consistent with the expectation that NMD targets are being actively degraded, we found that

NMD targets have lower mRNA abundances than other genes; this is in strong contrast to

TuORF mRNAs, which have higher mRNA abundances than other genes (**Figure 5A**). While

both NMD targets and TuORF genes are associated with lower TE, shorter mRNA half-lives,

and lower protein abundance than other genes (**Figure 5B-D**), the NMD targets are associated

with even lower protein levels than TuORF genes (**Figure 5D**).

Interestingly, while both NMD targets and TuORF mRNAs have lower translation

efficiency (**Figure 5B**), TuORF-containing NMD targets have significantly lower translation

efficiency than either NMD targets without TuORFs or TuORF-containing non-NMD target

genes (**Figure 5B**). This additive repression of translation efficiency for TuORF-containing NMD

targets suggests that NMD and TuORFs suppress translation through separate mechanisms.

Our observation is consistent with the expectation that TuORFs prevent the scanning pre-

initiation complex from reaching the mORF while NMD suppresses translation through UPF1

and/or other translational repressors (Isken et al., 2008; Raxwal et al., 2020; Zinshteyn et al.,

2021). Together, the differences in the expression patterns between TuORF mRNAs and NMD

targets further support the idea that TuORFs are not an mRNA feature that triggers NMD in

Arabidopsis.

TuORF mRNAs and NMD targets are degraded through different mRNA decay

mechanisms

We next investigated the mechanisms underlying the decay of TuORF mRNAs and NMD

targets. mRNA decay can occur in the 5' to 3' or 3' to 5' direction, which is carried out by

different machinery (Sieburth and Vincent, 2018). A genome-wide survey of mRNA decay rates

in wild type and mutants defective in either 5' to 3' decay or 3' to 5' decay, or both, has been

conducted in Arabidopsis (Sorenson et al., 2018). In the survey, the *vcs-7* mutant, a null mutant

of *VARICOSE*, which encodes a scaffold protein for the mRNA decapping complex, was used to

examine the 5' to 3' mRNA decay; the *sov* mutant, defective in *SUPPRESSOR OF VARICOSE*,

which encodes a ribonuclease II, was used to examine 3' to 5' decay; and the *vcs/sov* double

mutant was used to examine decay affected in both directions (Deyholos et al., 2003; Goeres et

al., 2007; Xu et al., 2006; Zhang et al., 2010). We exploited this published dataset (Sorenson et

al., 2018) to study the genetic requirements for the mRNA decay of TuORF mRNAs, NMD

targets, TuORF-containing NMD targets, and other genes (**Figure 6**).

Similar to mRNAs that are non-NMD targets and lack TuORFs (**Figure 6D, 6H**), TuORF

mRNAs had dramatically reduced decay rates in the *vcs* mutant and the *vcs/sov* double mutant

(**Figure 6A, 6E**), indicating that decapping and 5' to 3' decay are critical for degrading these

transcripts. In contrast, the mRNA decay of NMD targets without TuORFs was only significantly

lower in the *vcs/sov* double mutant, suggesting that NMD targets are efficiently degraded by

both the 5' to 3' and 3' to 5' pathways (**Figure 6B, 6F**). Although there are fewer NMD targets

containing TuORFs, which resulted in a small sample size and statistical insignificance (**Figure

6C, 6G**), the decay rates in this group of genes in the four genetic backgrounds showed trends

similar to those of NMD targets without TuORFs (**Figure 6B, 6F**).

Together, our results suggest that TuORF mRNAs and NMD targets have distinct

genetic requirements for mRNA decay (see model in **Figure 6I**). Although other mRNAs are

also decayed through the 5' to 3' pathway (**Figure 6D, 6H**), the TuORF mRNAs appear to

undergo accelerated decay as their mRNA half-lives are shorter (**Figure 2C, 6I**).

DISCUSSION

In this study, we identified Arabidopsis TuORFs using improved Ribo-seq data and

systematically examined their presumed roles in gene expression. While our results support the

expected roles of uORFs in repressing mORF translation and mRNA stability, our study reveals

distinct features and regulation of uORFs in plants compared with animals.

Our observations that plant TuORF genes have higher expression levels (**Figure 2**),

encode larger proteins (**Figure 3**), and are enriched for evolutionarily conserved domains

(**Table S2**) raised an interesting question as to whether older genes are more likely to be

controlled at the protein level through mRNA degradation and translation via uORFs. Higher

mRNA abundance may allow TuORF mRNAs to be highly translated under certain conditions

that require a high level of specific proteins. For example, it was suggested that derepression of

downstream translation is a general mechanism of uORF-mediated expression under stress

(Andreev et al., 2018). Thus, TuORFs could enable fast and dynamic regulation of mORF

translation. Future work monitoring the translation of TuORFs and their mORFs under different

conditions will help decipher the regulation and physiological roles of specific TuORFs.

In this report, we present five pieces of evidence that Arabidopsis TuORFs generally do

not trigger NMD. First, fewer than 3% of TuORF genes were identified in the high-confidence

NMD target list (**Tabls S4B**). Second, NMD targets, but not TuORF mRNAs, are upregulated in

various NMD mutants (**Figure 4**). Third, TuORF mRNAs have significantly higher transcript

levels, while the NMD targets have lower transcript levels compared with other genes (**Figure**

5A). Fourth, NMD and TuORFs additively repress translation (**Figure 5B**). Finally, while NMD

targets are efficiently degraded through both 5' to 3' and 3' to 5' decay, TuORF mRNAs rely on

5' to 3' decay like most other genes (**Figure 6**). It is unlikely that a large number of TuORF

transcripts escape NMD altogether; rather, our results suggest TuORFs generally do not trigger

NMD in Arabidopsis.

The fact that Arabidopsis is missing key NMD factors found in animals, including SMG1,

SMG5, and SMG6, suggests significant variations in the NMD mechanisms between plants and

animals. In particular, SMG1 is the kinase responsible for UPF1 phosphorylation required for

substrate recognition in animal NMD (He and Jacobson, 2015; Kurosaki et al., 2019). The lack

of SMG1 and use of an alternative kinase for Arabidopsis UPF1 phosphorylation (Kerényi et al.,

2013) may lead to a difference in substrate recognition: while TuORF mRNAs are recognized as

NMD targets in animals, plant TuORF mRNAs are not.

Despite the difference in NMD components in plants and animals, our findings that

Arabidopsis NMD targets are degraded through both 5' to 3' decay and 3' to 5' decay is

consistent with the two-directional decay of NMD targets in animals. Note that plant SOV is the

homolog of DIS3L2, the exonuclease responsible for 3' to 5' decay of NMD targets in animals

(Chang et al., 2013; Lubas et al., 2013). Two-directional decay could be advantageous to

quickly remove aberrant mRNAs from the cytoplasm to prevent further translation and facilitate

the recycling of NMD machinery.

Genes controlled by both NMD and TuORFs might encode critical regulators, and their

protein levels require precise regulation. These genes included key enzymes in gibberellin

hormone and polyamine biosynthesis and many transcription factors (**Figure 1F-G, and Table**

S4B). For example, one of the transcription factors, *SUPPRESSOR OF ACAULIS 51* (*SAC51*),

was discovered in a genetic screen for suppressors of the dwarf phenotype of the *acaulis5*

mutant (Imai et al., 2006). When the uORF in *SAC51* was mutated, it increased the transcript

level of *SAC51*, which encodes a bHLH transcription factor that was proposed to activate the

genes required for stem elongation.

It is well established that there is an important interplay between translation efficiency

and mRNA stability. Two mechanisms were proposed to explain this phenomenon: the "stalled

ribosome-triggered decay model," which predicts that ribosome stalling or slowing promotes

mRNA degradation, and the "translation factor-protected model," which predicts that active

translation promotes mRNA stability because the translation initiation factors protect the mRNA

from the decapping complex (Presnyak et al., 2015; Radhakrishnan et al., 2016; Beelman and

Parker, 1994; LaGrandeur and Parker, 1999; Schwartz and Parker, 2000; Chan et al., 2018).

Consistent with the first model, several CPuORF-containing mRNAs in Arabidopsis were

reported to have strong ribosome stalling patterns, and the endonucleolytic cleavages upstream

of the stalled ribosomes were observed (Hou et al., 2016; Yu et al., 2016). While most of the

TuORFs we identified did not display clear stalling patterns (e.g., **Figure 1F-G, S1B, S2A-N**), it

remains possible that ribosomes have lower translation speeds, on average, during uORF

translation. Because ribosomes will quickly approach termination after initiation due to the short

length of uORFs, and because both initiation and termination are major time-consuming steps in

translation (Shah et al., 2013; Lawson et al., 2021), the average translation speed per codon for

uORFs is likely to be slower than that of longer ORFs. Alternatively, if cells mainly monitor the

translation efficiency of the mORF, then the poorer mORF translation caused by uORFs could

explain the faster degradation of TuORF mRNAs predicted by this model. In contrast, since

ribosomes actively engage with TuORFs, the "translation factor-protected" model does not fit

TuORF mRNAs, especially because their degradation relies on decapping and 5' to 3' decay in

plants. Further investigation into TuORF-mediated mRNA degradation will help elucidate the

role of uORFs in regulating mRNA stability.

MATERIALS AND METHODS

Plant growth conditions and lysate preparation

Arabidopsis Col-0 seeds were surface sterilized with 70% ethanol for 5 min, followed by 33%

bleach and 0.03% Tween 20 for 10 min, then rinsed with sterile water 5 times. The seeds were

imbibed at 4°C in the dark for 2 days, then grown hydroponically in sterile liquid media (2.15 g/L

Murashige and Skoog salt, 1% sucrose, 0.5 g/L MES, pH 5.7) while shaking at 85 rpm under a

16-h light (75–80 µmol m^{-2}·s^{-1} from cool white fluorescent bulbs) and 8-h dark cycle at 22°C for

7 days. At Zeitgeber time 4 (4 h after lights on), DMSO corresponding to 0.1% of the media

volume was added to the media (these were mock samples of our large-scale experiment). After

20 and 60 min, three biological replicates (~300 plants per sample) were harvested at each time

point and immediately flash frozen in liquid nitrogen.

Plant lysates were prepared as previously described (Hsu et al., 2016). Briefly, 0.1 g of ground

tissue power was resuspended in 400 µL of lysis buffer (100 mM Tris-HCl [pH 8], 40 mM KCl,

20 mM MgCl2, 2% [v/v] polyoxyethylene [10] tridecyl ether [Sigma, P2393], 1% [w/v] sodium

deoxycholate [Sigma, D6750], 1 mM dithiothreitol, 100 µg/mL cycloheximide [Sigma, C4859],

100 µg/mL chloramphenicol [Sigma R4408], and 10 units/mL DNase I [Epicenter, D9905K]).

The lysates were spun at 3,000 g for 3 min, and the supernatant was transferred to a new tube

and subsequently centrifuged at 20,000 g for 10 min. The supernatant was transferred to a new

tube and the RNA concentration was determined with 10x dilutions using the Qubit RNA HS

assay (Thermo Fisher Scientific; Q32852). Aliquots of 100 µL and 200 µL of the lysates were

made, and they were flash frozen in liquid nitrogen and stored at -80°C until further processing.

Ribo-seq library construction

Ribosome footprints were obtained using 200 µL of the lysates described above, and

sequencing libraries were constructed according to our previous method (Hsu et al., 2016) with

the following modifications: briefly, after RNase I digestion (50 units nuclease per 40 µg of RNA;

the nuclease was included in TruSeq Mammalian Ribo Profile Kit, Illumina, RPHMR12126) and

passing through a size exclusion column (Illustra MicroSpin S-400 HR Columns; GE Healthcare;

27-5140-01), RNA > 17 nt was isolated with the RNA Clean & Concentrator-25 kit (Zymo

Research, R1017) and separated on 15% urea-TBE gels (Invitrogen, EC68852BOX). Gel slices

between 28 and 30 nt were isolated, and the RNAs were purified as previously described. Next,

rRNA depletion was performed using RiboZero Plant Leaf kit (Illumina, MRZPL1224) in one

quarter of the recommended reaction volume. Ribo-seq libraries were then constructed using

the TruSeq Mammalian Ribo Profile Kit (illumina, RPHMR12126) as previously described with

nine cycles of PCR amplification. Libraries with equal molarity were pooled and sequenced on a

Hi-Seq 4000 sequencer using single-end 50-bp sequencing.

RNA-seq library construction

Total RNA greater than 200 nt was purified from 100 µL of the lysates described above using

the RNA Clean & Concentrator-25 kit (Zymo Research, R1017) as previously described (Hsu et

al., 2016). RNA integrity was evaluated using a Bioanalyzer (Agilent) RNA pico chip, and RNA

integrity numbers (RINs) ranging from 7.2 to 7.7 were observed among the samples. A total of 4

µg of RNA per sample was subjected to rRNA depletion using the RiboZero Plant Leaf kit

(Illumina, MRZPL1224) following the manufacturer's recommendations. Then, 100 ng of rRNA-

depleted RNA was fragmented to around 200-nt long based on the RIN reported by the

Bioanalyzer, and strand-specific sequencing libraries were made using the NEBNext Ultra II

Directional RNA Library Prep Kit (New England Biolabs, E7760S) with eight cycles of

amplification. Libraries of equal molarity were pooled and sequenced on a Hi-Seq 4000 using

paired-end 100-bp sequencing.

Data pre-processing and analysis

Data pre-processing and analysis were performed similarly to that previously described (Hsu et

al., 2016), except that the Araport11 annotation was used in this study. Briefly, for Ribo-seq

libraries, the adaptor (AGATCGGAAGAGCACACGTCT) was clipped with fastx_clipper (FASTX

toolkit v0.0.14) (http://hannonlab.cshl.edu/fastx_tool- kit/). For both RNA-seq and Ribo-seq, we

used Bowtie2 (v2.3.4.1) (Langmead and Salzberg, 2012) to remove

rRNA/tRNA/snRNA/snoRNA sequences. Both the RNA-seq and Ribo-seq reads were mapped

to the transcriptome with STAR aligner (Dobin et al., 2013) (RNA-seq parameters: --

outFilterMismatchNmax 2 --outFilterMultimapNmax 20 --outFilterType BySJout --

alignSJoverhangMin 8 --alignSJDBoverhangMin 2; all parameters used for Ribo-seq were the

same except --outFilterMismatchNmax 1). Combining both 20-min and 60-min samples, the

total reads mapped to the genome for Ribo-seq and RNA-seq were 298.01 million and 180.67

million, respectively. We then used the bam files from both RNA-seq and Ribo-seq for

RiboTaper (Calviello et al., 2016) to identify translated ORFs. The Ribo-seq metaplot was

created using the create_metaplots.bash function in RiboTaper (v1.3.1a). The distribution of

Ribo-seq reads in different genome features was calculated using Ribo-seQC. The Ribo-seq

read lengths and offsets for RiboTaper were 24, 25, 26, 27, 28 and 8, 9, 10, 11, 12,

respectively. The discovered uORFs and mORFs were extracted from the RiboTaper output

ORF_max_filt file. To calculate TE, we first used STAR to map the RNA-seq and Ribo-seq

reads to the CDS of annotated coding genes. The resulting bam files were used to quantify the

transcripts per million (TPM) of each gene with RSEM (v1.3.1) (Li and Dewey, 2011). Then,

translation efficiency was calculated by dividing the Ribo-seq TPM by the RNA-seq TPM. The

root and shoot data from Hsu et al., 2016 (Hsu et al., 2016) were reanalyzed using the

Araport11 annotation. The P_sites_all files from RiboTaper were processed using the following

code: cut -f 1,3,6 P_sites_all | sort | uniq -c | sed -r 's/^(*[^]+) +/\1\t/' > output.txt to combine the

read counts at each P-site. Data visualization and statistical analysis were performed in R

(v4.0.3) (R Core Team (2013), 2017). In particular, the gene levels from the Ribo-seq and RNA-

seq reads were visualized with RiboPlotR (Wu and Hsu, 2021).

The mRNA half-life and mRNA-decay rate data were downloaded from Szabo et al. and

Sorenson et al., respectively (Szabo et al., 2020; Sorenson et al., 2018). The quantitative

proteomics data for Arabidopsis shoots and roots were downloaded from Song et al. (Song et

al., 2018). The Arabidopsis gene expression levels (TPM) used for investigating mRNA levels

for uORF genes in different Arabidopsis organs, growth stages, and ecotypes were directly

downloaded from the EMBO-EBI expression atlas (E-MTAB-7978: Arabidopsis tissue atlas, E-

GEOD-53197: 17 Arabidopsis thaliana accessions, E-MTAB-4812: tomato root, leaf, flower [two

stages], and fruit [six stages], and E-MTAB-4813: three longitudinal sections of six stages during

tomato fruit development) (Papatheodorou et al., 2020). Our previous Ribo-seq and RNA-seq

data for Arabidopsis roots and shoots were described in (Hsu et al., 2016); our previous Ribo-

seq and RNA-seq data for tomato roots were described in (Wu et al., 2019). The published

RNA-seq data for the NMD mutants *upf1* (*lba1*) *upf3-1* and its wild-type control (GEO:

GSE41432), *smg7-1 pad4* and its *pad4* control (GEO: GSE55884), *upf1-3 pad4* and its *pad4*

control (GEO: GSE146340), and *upf2-10* and its wild-type control (SRA: SRP056795) were

downloaded, and the mRNA levels were quantified by Kalisto (Bray et al., 2016) and the TPM

values were further analyzed to compare the steady-state mRNA levels for different categories

of genes.

DATA ACCESS

All raw and processed sequencing data generated in this study have been submitted to the

NCBI Gene Expression Omnibus (GEO; https://www.ncbi.nlm.nih.gov/geo/) under accession

number GSE183264.

COMPETING INTEREST STATEMENT

The authors declare that there is no conflict of interest.

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AUTHOR CONTRIBUTIONS

HLW and PYH designed the research, PYH performed the sequencing experiments, HLW

analyzed the sequencing data, HLW and PYH interpreted the data, and HLW and PYH wrote

the paper.

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FIGURE and FIGURE LEGENDS

Figure 1

Figure 1. Enhanced Ribo-seq coverage improves uORF identification

(A) Metagene analysis of 28-nt ribosome footprints mapped to regions near the start and

stop codons of annotated ORFs. The reads are presented with their first nt at the P-site,

which is the 13th nt for 28-nt footprints. The reads are colored in red, blue, and green to

indicate they are in the first (expected), second, and third reading frames, respectively.

The majority of footprints were mapped to the CDS in the expected reading frame

(92.4% in frame).

(B) Genomic features mapped based on ribosome footprints. Reads that mapped to nuclear

(Nuc)-, mitochondrial (Mt)-, and plastid (Pt)-encoded genes are shown.

(C) Length distribution of ribosome footprints. Reads that mapped to nuclear-encoded genes

are presented.

(D) Distinct P-sites detected in 1 to 20 million randomly selected ribosome footprints from

our current and previous datasets.

(E) Numbers of TuORFs, CPuORFs, and mORFs identified in our current and previous

datasets.

(F-G) RNA-seq and Ribo-seq profiles of *ATKAO1 and SAMDC1* from our current data. RNA-

seq coverage is shown with a light-yellow background. Ribo-seq reads are presented

with their first nt at the P-site, and they are colored in red, blue, and green to indicate

they are in the first (expected), second, and third reading frames, respectively. Within the

gene models, yellow boxes represent the TuORFs, black boxes represent the annotated

mORFs, and gray and white regions indicate 5' UTRs and 3' UTRs, respectively. The

specific isoform being considered is indicated to the left of the gene model. Within the

profiles, green and orange vertical dashed lines represent translation start and stop,

respectively, for the TuORF. Black and gray vertical dashed lines represent translation

start and stop, respectively, for the annotated mORF.

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Figure 2

Figure 2. TuORF-containing genes have higher mRNA abundance but are associated with lower TE, shorter mRNA half-lives, and lower protein abundance

(A-D) The mRNA levels, translation efficiencies, mRNA half-lives, and protein levels of TuORF, UuORF, and no-uORF genes in Arabidopsis seedlings. The mRNA half-life data, which were measured with metabolic labeling in Arabidopsis seedlings, were extracted from (Szabo et al., 2020), and the protein abundances as determined by quantitative proteomics in Arabidopsis seedling shoots were extracted from (Song et al., 2018). Within the boxplots, the dashed lines mark the median level of the no-uORF genes; the number of genes in each category is listed below the lower whisker. The ratios underneath the plots indicate the median of each group normalized to that of the no-uORF genes. The statistical significance was determined by Wilcoxon signed-rank test. The adjusted p-value was determined using the Benjamini and Yekutieli procedure to control for the false discovery rate in multiple testing (* $0.05 > p > 0.01$, ** $0.01 > p > 0.001$, *** $0.001 > p > 1e-4$, **** $1e-4 > p > 0$). These statistical analyses and graphic layouts are the same for all boxplots in the study.

(E-F) The mRNA levels and translation efficiencies of TuORF, UuORF, and no-uORF genes from reanalysis of our previous tomato root data (Wu et al., 2019). The dashed lines mark the median level of the no-uORF genes.

(G) The mRNA levels of TuORF, UuORF, and no-uORF genes in 46 Arabidopsis tissues and developmental stages from independent studies. The TPM values from RNA-seq quantification were downloaded from the EMBO-EBI expression atlas (Papatheodorou et al., 2020).

(H) The distributions of the mRNA levels of TuORF and UuORF genes in Arabidopsis seedlings. Vertical dashed lines indicate the median value of each group.

Figure 3

Ratio: 2.53 2.04 1   1.33 1.04 1   1.10 1.05 1   1.53 1.24 1

Figure 3. TuORF genes are larger than other genes

Lengths of the 5' UTRs (A), CDSs (B), 3' UTRs (C), and entire transcripts (D) of TuORF,

UuORF, and no-uORF genes. Within the boxplots, the dashed lines mark the median level of

the no-uORF genes; the number of genes in each category is listed below the lower whisker.

The ratios underneath the plots indicate the median of each group normalized to that of the no-

uORF genes. The statistical significance was determined as described in the Figure 2 legend.

Figure 4

Figure 4: TuORF mRNAs are not upregulated in the NMD mutants

The ratios of mRNA levels between NMD mutants and their control plants from four RNA-seq

datasets (Drechsel et al., 2013; Merchante et al., 2015; Raxwal et al., 2020; Gloggnitzer et al.,

2014) were used to examine the effects of the NMD pathway on NMD targets (NMDt) and

TuORF genes. (A) *upf1* (*lba1* allele) *upf3-1* double mutant compared with wild type (Col-0). (B)

upf2-10 compared with wild type (Col-0). (C) *upf1-3 pad4* compared with *pad4*. (D) *smg7-1*

pad4 compared to *pad4*. See Table S3 for specific growth conditions for each dataset. Within

the boxplots, the dashed lines mark the median level of other mRNAs; the number of genes in

each category is listed below the lower whisker. The ratios underneath the plots indicate the

median of each group normalized to that of other mRNAs. The statistical significance was

determined as described in the Figure 2 legend.

Figure 5

Figure 5. Expression patterns of NMD targets and TuORF genes.

(A-D) The mRNA levels (A), translation efficiencies (B), mRNA half-lives (C), and protein levels (D) of NMD targets (NMDt) without or with TuORFs compared with those of TuORF mRNAs and other mRNAs. Within the boxplots, the dashed lines mark the median level of other mRNAs; the number of genes in each category is listed below the lower whisker. The ratios underneath the plots indicate the median of each group normalized to that of other mRNAs. The statistical significance was determined as described in the Figure 2 legend.

Figure 6

Figure 6. TuORF mRNAs and NMD targets are degraded through different mRNA decay mechanisms

Cumulative plots (A-D) and boxplots (E-H) showing the mRNA decay rates for NMD targets and TuORF genes in wild type and mutants defective in either 5' to 3' decay (*vcs-7*), 3' to 5' decay (*sov*), or both (*vcs sov).* The mRNA decay rates were extracted from (Sorenson et al., 2018). Because Col-0 is naturally an *sov* mutant, the wild type here is Col-0 carrying the functional *SOV* allele from Landsberg *erecta* driven by its native promoter (Zhang et al., 2010; Goeres et al., 2007). The *sov* mutant here is Col-0, and *vcs* is Col-0 with the *vsc-7* allele and the L*er SOV* transgene. The *vcs sov* double mutant is Col-0 carrying the *vcs-7* mutation (Sorenson et al.,

2018). Within the boxplots, the dashed lines mark the median level of wild type; the number of

genes in each category is listed below the lower whisker. The ratios underneath the plots

indicate the median of each group normalized to that of wild type. The statistical significance

was determined as described in the Figure 2 legend. (I) Model of mRNA decay for NMD targets,

TuORF mRNAs, and other mRNAs. Orange packman indicates 5'-to-3' decay machinery and

hot pink packman indicates 3'-to-5' decay machinery.

SUPPLEMENTARY DATA

Figure S1. Correlations among samples and comparisons between our previous and current

datasets

Figure S2. Examples of TuORFs in important regulatory genes.

Figure S3. mRNA levels of uORF-containing genes in different ecotypes/cultivars, growth

stages, tissues in Arabidopsis and tomato

Figure S4. mRNA levels of CPuORF-containing genes compared to no-uORF genes.

Figure S5. Protein abundance of the uORF-containing genes in Arabidopsis root.

Figure S6. Distributions of uORF peptide lengths in NMD target and non-NMD target TuORF

genes.

Figure S7. Length of UTRs and CDS for NMD targets and TuORF genes

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Table S1. Translated ORFs identified by RiboTaper (Excel)

Table S1A-C. Translated uORFs identified in the current dataset, our previous Arabidopsis

shoot and root datasets

Table S1D-F. Translated mORFs identified in the current dataset, our previous Arabidopsis

shoot and root datasets

Table S2. GO-term analsyis of TuORF genes (Excel)

Table S3. Summary of the large-scale datasets used in this study

Table S4. NMD target gene list (Excel)

Table S4A. Overlap between Raxwal et al. high-confidence NMD targets and our TuORF genes

Table S4B. 49 genes identified as high-confidence NMD target and TuORF genes

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1017 **SUPPLEMENTARY FIGURES**

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Figure S1. Correlations among samples and comparisons between our previous and

current RNA-seq and Ribo-seq datasets. (A) Pearson correlation among three biological

replicates of our current RNA-seq and Ribo-seq analyses of samples harvested after 20 or 60

min of DMSO treatment (these are control samples of our large-scale experiment). TPM values

in the CDSs of annotated protein-coding genes were used. (B) Higher Ribo-seq read coverage

was obtained in both the uORF and mORF regions in our current dataset; *RGA1* is shown as an

example. Note that the mRNA levels are similar between our previous data (Hsu et al. 2016,

shoot) and current data. Data representations are the same as those described in the Figure 1F

legend.

Figure S2. TuORFs in known important regulatory genes

A. *TOC1*

B. *SCARECROW*

C. *ARF2*

D. *ARF3*

E. *ARF8*

E. *ARF8* (continued)

F. *Phytochrome B*

G. *Phytochrome E*

H. *PIF3*

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Figure S2. Examples of TuORFs in important regulatory genes. Data are presented as

described in the Figure 1F legend. Isoform numbers are shown to the left of the gene models,

and <u>the isoform being considered is bolded</u>. For genes that have multiple TuORFs identified (B,

E, and I), the 3-nt periodicity of each TuORF is presented in a separate panel.

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Figure S3. mRNA levels of uORF-containing genes in different growth stages and tissues

of various Arabidopsis ecotypes and tomato.

(A) Arabidopsis roots and aerial and floral parts from different ecotypes.

(B) Various developmental stages of tomato fruit (Heinz 1706 cultivar).

(C) Roots, shoots, flowers, and fruits of tomato (Heinz 1706 cultivar).

The TPM values from RNA-seq quantification were extracted from the EMBO-EBI expression

atlas (Papatheodorou et al. 2020) (see MATERIALS AND METHODS).

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Figure S4. mRNA levels of CPuORF-containing genes compared with no-uORF genes.

Within the boxplot, the dashed line marks the median level of no-uORF genes; the number of

genes in each category is listed below the lower whisker. The ratios underneath the plots

indicate the median of each group normalized to that of no-uORF genes. The statistical

significance was determined by Wilcoxon signed-rank test. The adjusted p-value was

determined using the Benjamini and Yekutieli procedure to control for the false discovery rate in

multiple testing (* 0.05 > p > 0.01, ** 0.01 > p > 0.001, *** 0.001 > p > 1e-4, **** 1e-4 > p > 0).

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Figure S5. Protein abundance of the uORF-containing genes in Arabidopsis roots. The

protein quantification data from Arabidopsis seedling roots were extracted from Song et al.

(2018). Within the boxplot, the dashed line marks the median level of no-uORF genes; the

number of genes in each category is listed below the lower whisker. The ratios underneath the

plots indicate the median of each group normalized to that of no-uORF genes. The statistical

significance was determined by Wilcoxon signed-rank test. The adjusted p-value was

determined using the Benjamini and Yekutieli procedure to control for the false discovery rate in

multiple testing (* 0.05 > p > 0.01, ** 0.01 > p > 0.001, *** 0.001 > p > 1e-4, **** 1e-4 > p > 0).

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1069   **Figure S6.** Distributions of uORF peptide lengths in NMD target and non-NMD target TuORF

1070   genes. Vertical dashed lines indicate the median value of each group.

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Figure S7. Lengths of UTRs and CDSs for NMD targets and TuORF genes.

Lengths of the 5' UTRs (A), CDSs (B), 3' UTRs (C), and entire transcripts (D) of NMD targets

without a TuORF, NMD targets with a TuORF, non-NMD TuORF genes, and other genes.

Within the boxplots, the dashed lines mark the median level of other mRNAs; the number of

genes in each category is listed below the lower whisker. The ratios underneath the plots

indicate the median of each group normalized to that of other mRNAs. The statistical

significance was determined by Wilcoxon signed-rank test. The adjusted p-value was

determined using the Benjamini and Yekutieli procedure to control for the false discovery rate in

multiple testing (* $0.05 > p > 0.01$, ** $0.01 > p > 0.001$, *** $0.001 > p > 1e-4$, **** $1e-4 > p > 0$).

Table S3. Summary of datasets used in this study

Reference	Dataset	Sample	Genotype	Growth conditions
Current study	Current study RNA-seq Ribo-seq	7-day-old Arabidopsis seedlings	Col	16h-light/8h-dark, grown in 0.5x MS liquid media, collected between 4–5 h after light on
(Hsu et al., 2016)	Our previous data RNA-seq Ribo-seq	4-day-old Arabidopsis seedling shoots or roots	Col	16h-light/8h-dark, grown in 0.5x MS liquid media, collected at 3 h after light on
(Song et al., 2018)	Our previous quantitative proteomics data	4-day-old Arabidopsis seedling shoots or roots	Col	16h-light/8h-dark, grown in 0.5x MS liquid media, collected at 3 h after light on
(Szabo et al., 2020)	mRNA half-life (metabolic labeling)	5-day-old Arabidopsis seedlings	Col	Constant light, grown in 0.5x MS liquid media
(Wu et al., 2019)	Our previous tomato data RNA-seq Ribo-seq	4-day-old tomato seedling roots	Heinz 1706	16h-light/8h-dark, grown on 1x MS solid media, collected at 3 h after light on
(Drechsel et al., 2013)	NMD mutant RNA-seq 1	11-day-old Arabidopsis seedlings	Col, *upf1 (lba1)* *upf3-1*	Constant light, grown on 0.5x MS solid media
(Merchante et al., 2015)	NMD mutant RNA-seq 2	3-day-old Arabidopsis etiolated seedlings	Col, *upf2-10*	Darkness, grown on 1x MS solid media
(Raxwal et al., 2020)	NMD mutant RNA-seq 3	Arabidopsis 8-week-old leaves for *upf1 pad4*, 3–4-week-old leaves for *pad4*	*upf1-3 pad4*, *pad4*	16h-light/8h-dark, soil
(Gloggnitzer et al., 2014)	NMD mutant RNA-seq 4	Arabidopsis likely 4–5-week-old leaves	*smg7-1 pad4*, *pad4*	16h-light/8h-dark, soil
(Sorenson et al., 2018)	mRNA decay rate	5-day-old Arabidopsis seedlings	Wild type, *vcs*, *sov*, *vcs sov*	Constant light, grown on 0.5x MS solid media, treated with liquid solution