Mettl5 coordinates protein production and degradation of PERIOD to regulate sleep in

Drosophila

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Summary

Sleep plays a critical role in animal physiology, primarily governed by the brain, and its disruption is

prevalent in various brain disorders. Mettl5 is associated with intellectual disability (ID), which often

includes sleep disturbances. However, the mechanism underlying these sleep disruptions in ID remains

poorly understood. In this study, we investigated the sleep phenotypes resulting from *Drosophila*

Mettl5 mutations. Rescue experiments revealed that *Mettl5* functions predominantly within neurons

and glia marked by *Mettl5*-Gal4 to regulate sleep. Previous work established that Mettl5 forms a

complex with Trmt112 to influence rRNA methylation. Notably, a mutation in *Trmt112* mutation

recapitulated these sleep disturbances, implicating translational regulation by the Mettl5/Trmt112

complex. Subsequent RNA-seq and Ribo-seq analyses of *Mettl5^{1bp}* mutants uncovered downstream

effects, including altered expression of proteasome components and clock genes. Rescue experiments

confirmed that the net increase in PERIOD protein underlies the sleep phenotype. This study

illuminates the interplay between ribosome function, clock genes, and the proteasome in sleep

regulation, highlighting the integrated roles of protein synthesis and degradation. These findings could

potentially provide an example for *in vivo* study of rRNA methylation function, expand our

understanding of protein homeostasis in sleep and offer insights into the sleep phenotypes associated

with ID.

Key words: *Mettl5*, rRNA methyltransferase, proteasome, Clock genes, sleep

Introduction

Sleep is essential for animal physiology, and understanding its molecular mechanisms has significant implications for both basic research and clinical applications. Although previous studies have identified key sleep regulators (*Du et al., 2021*), fundamental questions remain regarding the roles of cellular processes such as protein synthesis and degradation in sleep regulation.

Growing evidence suggests a dynamic interplay between sleep regulation and protein homeostasis. Protein synthesis is particularly active during sleep (*Lyons et al., 2023*), while Sleep deprivation affects translational initiation (*Costa et al., 2019*). The proteasome, a critical mediator of ubiquitin-dependent protein degradation, regulates synapse homeostasis by modulating ribosomal components (*Costa et al., 2019*), thereby influencing synaptic function. Intriguingly, proteasome components exhibit oscillating expression pattern at the transcriptional level in human cells (*Desvergne et al., 2016*). A recent study in *Drosophila* showed that mutation in a proteasome component alters sleep patterns (*Fernández et al., 2020*). Despite these advances, the precise mechanisms by which protein synthesis and degradation influence sleep, and how these processes interact, remain unresolved.

While several factors influencing proteasome assembly and subunit transcription have been identified (*Kapetanou et al., 2022*), the translational control of proteasome subunits remains poorly understood. Additionally, proteasome inhibition has been shown to alter ribosome function (*Galimberti et al., 2016; Costa et sl., 2019; Palanca et al., 2014*), raising the question of whether the proteasome itself is regulated by the status of the translation machinery. Exploring this relationship could provide key insights into the interplay between proteasome activity and translation.

In *Drosophila*, *Mettl5* exhibits 18S ribosomal RNA m^{6}A methyltransferase activity (*Leismann et al., 2020*) and interacts with Trmt112 to facilitate this function, a mechanism conserved in human (*Leismann et al., 2020; van Tran et al., 2019*). Although *Mettl5* depletion abolishes m^{6}A modification on 18S rRNA without impairing rRNA maturation, it affects fly orientation behavior. In mammalian cells, *Mettl5*-mediated 18S rRNA N6-methyladenosine (m^{6}A) modification regulates stem cell fate determination and neural function (*Wang et al.*, *2020*). However, it remains unclear whether *Mettl5* modulates global or transcript-specific translation profile *in vivo*, particularly in tissues rather than cell lines. Moreover, while *Mettl5* is linked to intellectual disability (ID) with comorbid sleep disturbances,

its mechanistic role in these disorders is unknown. Testing whether *Mettl5* loss affects behaviors like sleep in *Drosophila* could clarify this connection.

We discovered that *Mettl5* regulates sleep in *Drosophila*. To investigate the underlying mechanism, we performed RNA-seq and Ribo-seq on *Mettl5* mutants, revealing dysregulation of multiple clock genes and proteasome components. This suggests *Mettl5* coordinates protein production and degradation, which are crucial for protein homeostasis. Follow-up experiments confirmed that the protein level of Period was upregulated in *Mettl5* mutants, contributing to the sleep phenotype. This study highlights that ribosome defects can perturb proteasome function, uncovering a mechanism that couples protein degradation with synthesis. Additionally, by mapping the genome-wide downstream gene profile of an rRNA methylation modifier, this study offers insights into the gene-specific roles of ribosome function.

Results

Mettl5* regulates sleep in *Drosophila

Mettl5 contains a predicted N6 adenine specific nucleic acids methyltransferase domain. In order to study the function of this gene in sleep regulation, we generated CRISPR-Cas9 knockout mutants. We created two alleles. *Mettl5^{1bp}*, which deleted 1bp in the coding region results in a truncated version (Figure 1A) and down regulation of *Mettl5* at mRNA level (Figure 1B). Another allele *Mettl5^{9bp}*, which results in 3 amino acids deleted version of *Mettl5* (Figure 1A) does not cause significant change at mRNA level (Figure 1B). We found that heterozygous *Mettl5^{1bp}* mutants exhibited significantly reduced nighttime sleep (Figure 1C-G), particularly during early night (ZT12-ZT16; Figure 1C). Quantitative analysis revealed increased wakefulness at ZT14 in mutants (Figure 1H). 'awake %' was used to indicate the percentage of awake fruit fly population at specific time points (e.g., ZT14). quantitative nighttime sleep latency measurements indicated a delayed sleep start in mutants (Figure S1G). In addition, *Mettl5* expression was significantly up regulated during the sleep recovery period after mechanical sleep deprivation (Figure 1I-J). *Mettl5* mutant displayed significantly increased sleep rebound in 24 hours after sleep deprivation (Figure 1K-L), indicating its effects on sleep homeostasis. Moreover, results of sleep arousal assay at ZT19 indicated that the percentage of aroused flies are significantly more than the control group (Figure 1M). These results demonstrate that the truncated

Mettl5 mutation causes sleep deficits, establishing *Mettl5* as a novel sleep regulator in *Drosophila*.

To confirm the specificity of the sleep phenotype in *Mettl5^(1bp)*, we performed genetic rescue experiments. Introduction of a single wild-type *Mettl5* copy completely rescued both the reduced sleep amount and increased wakefulness at ZT14 (Figure 1N-R), demonstrating that these phenotypes specifically result from *Mettl5* deficiency.

The observed expression pattern of *Mettl5* further supports its sleep regulatory function. Using *Mettl5*-Gal4 reporter lines, we found expression in both neurons (colocalizing with ELAV staining; Figure S1A-C) and glial cells (colocalizing with Repo staining; Figure S1D-F). Behavioral analyses revealed additional mutant phenotypes consistent with sleep dysregulation. we tested the climbing ability of *Mettl5* mutation which showed an increase (Figure S1H). RNAi knocking down of *Mettl5* showed a consistent phenotype of down regulated sleep amount during the night time (Figure S1I-L).

Mettl5 regulates *Drosophila* sleep through its methyltransferase activity

Previous studies established that Mettl5 interacts with Trmt112 to regulate 18S rRNA m⁶A modification (*Leismann et al., 2020; van Tran et al., 2019*). Consistent with these findings, our LC-MS/MS analysis revealed significantly reduced m⁶A levels in both total RNA and 18S rRNA from heterozygous *Mettl5^(1bp)* (Figure 2A-B). To determine whether Mettl5's sleep regulatory function depends on its methyltransferase activity, we performed two key experiments. First, we examined the phenotype of *Trmt112* knockdown. We found that the *Mettl5*-Gal4 driven *Trmt112* RNAi resulted in a similar phenotype to *Mettl5^(1bp)* (Figure 2C-G), suggesting that *Mettl5* regulates sleep through its 18S rRNA m⁶A modification activity. The efficiency of the *Trmt112* RNAi line was tested in previous study (*López-Varea et al., 2021*). More important, we performed a rescue experiment with a mutated form of *Mettl5* that lacks the NPPF amino acids required for its methyltransferase activity (Figure2A) (Iyer et al., 2016). We observed that this mutant *Mettl5* failed to rescue the sleep phenotype caused by *Mettl5^(1bp)* (Figure 2H-L). Together, these results demonstrate that Mettl5's methyltransferase activity is essential for its role in sleep regulation, likely through its function in 18S rRNA m⁶A modification.

RNA-seq and Ribo-seq revealed the downstream gene profile of *Mettl5*

To better understand the downstream events of *Mettl5*, we performed RNA-seq and Ribo-seq to assess transcriptomic and translational changes in *Mettl5^(1bp)*. Principal Coordinates Analysis (PCoA)

of RNA-seq and Ribo-seq datasets revealed clear separation between mutant and control groups (Figure S2A, B, D, E). High reproducibility was found among biological replicates (Figure S2C, F), confirming data quality. Transcriptome analysis identified 1,053 significantly differentially expressed genes ($|\log_2$ (Fold Change) $| \geq 1$ & $p^{adj} < 0.05$), comprising 217 upregulated and 836 downregulated transcripts (Figure 3A, B). Parallel ribosome profiling revealed 299 translationally regulated genes ($|\log_2$ (Fold Change) $| \geq 0.265$ & $p^{adj} < 0.05$), with 149 upregulated and 150 downregulated targets compared to the controls (Figure 3C). Heatmaps of the top 100 differentially expressed genes from both RNA-seq and Ribo-seq analyses were generated according to the rlog transformed values (Figure S3A, B).

To explore the biological implication of differentially expressed genes at transcriptional and translational levels, we performed Gene Ontology (GO) enrichment analysis. At the transcriptional level (Figure S4A, C), the most significantly enriched biological processes included cellular response to chemical stimulus, small molecule biosynthetic process, and fatty acid metabolic process (Figure S4A), with supporting gene networks (Figure S4C). Translational-level analysis (Figure S4B, D) revealed prominent enrichment for organic acid and amino acid metabolic processes, lipid catabolism, cellular respiration, and transport mechanisms. Notably, we identified strong associations with circadian regulation, including circadian regulation of gene expression and entrainment of the circadian clock.

The GO results usually contain a long list of enriched terms that have highly redundant information and are difficult to summarize. So, we performed a simplify Enrichment analysis that visualizes the summaries of clusters by word cloud for the GO enrichment result at transcriptional and translational levels. For transcriptionally significant DEGs (Figure S5A), major enriched processes included metabolic and stimulus-response pathways, wing disc and imaginal system development, ion-channel homeostasis and transport, and cytokinesis/cell cycle transitions. Translationally regulated genes additionally showed enrichment for sleep-wake cycles, circadian behavior, and neuronal cell death (Figure S5D). Cellular component analysis highlighted chromosomal structures, membrane-bound complexes, and vesicles (Figure S5B), while molecular functions predominantly involved enzymatic activities, ion/protein binding, receptor activities, and transcription processes (Figure S5C).

The strong concordance between RNA-seq and Ribo-seq enrichment profiles (Figure S5E-F) further

validated these findings.

Our Gene Set Enrichment Analysis (GSEA) identified pathways enriched in differentially

expressed genes at both transcriptional and translational levels. To capture subtle but biologically

important changes that might be excluded by stringent statistical thresholds, we analyzed gene sets

ranked by log2 fold-change without p-value filtering. The GSEA-GO and GSEA-KEGG results, sorted

by Normalized Enrichment Score (NES), revealed several key pathways (Figure S6A-D). Notably, the

proteasome pathway showed significant suppression at both transcriptional and translational

levels (P.adjust<0.05) (Figure 3E, I). Transcriptional-level analysis highlighted disruptions in arginine

and proline metabolism and insect hormone biosynthesis (P.adjust<0.05) (Figure 3D, G). While, while

translational changes predominantly affected fatty acid degradation and phosphonate/phosphinate

metabolism (P.adjust<0.05) (Figure 3H-K).

To better understand the correlation between transcription and translation levels, we performed

correlation analysis of differentially expressed genes based on fold change of DEGs (Figure 3L). 977

genes showed opposing trends between transcription and translation, while only 24 genes exhibited

concordant changes. We identified 240 translation-specific and 3,309 transcription-specific DEGs,

with each cluster undergoing separate enrichment analysis (Figure S7A-H). Notably, core clock genes

(*Clk*, *tim*, etc.) displayed significant changes at both levels (Figure 3L).

To isolate translational effects, we calculated translation efficiency (TE) differences between

Mettl5^{1bp} and *w*^{1118} controls (Figure 3M). Among 1,204 genes with significantly altered TE, GO

enrichment highlighted amino acid metabolism and small molecule biosynthesis (Figure 3N), while

KEGG analysis emphasized glycine/serine/threonine metabolism and one-carbon pool by folate

pathways (Figure 3O, P).

Ribo-seq revealed the *Mettl5*^{1bp} led to changes of some global translation features

Using Ribo-seq, we compared global translation features between the *w*^{1118} and *Mettl5*^{1b} groups.

The TE (translation efficiency) correlation coefficients among the three biological replicates of each

genotype ranged from 0.87 to 0.93, demonstrating high reproducibility within genotypes (Figure S8A-

F). We found that the length of ribosome-protected RNA fragments (RPFs) was approximately 28nt

(Figure S10A). The RPFs exhibited a significant 3-nt periodicity. Metagene analysis of individual 28nt

reads revealed the distribution of RPFs across the gene locus in w^{1118} and $Mettl5^{1bp}$, respectively (Figure

S8, S9). The starting point of translation is 12nt upstream of start-codon and gradually disappears 15nt

from stop-codon (Figure S8G-L). RPFs on the metagene plot distribution around the translation start

and translation stop site (Figure S9A-F), different coding frames on CDS, 3'-UTR and 5'-UTR,

respectively in two groups of samples all showed the periodicity (Figure S9G-L). As expected, all

plots show an enrichment of P-sites in the first frame on the coding sequence but not the UTRs, in

accord with ribosome protected fragments from protein coding mRNAs (Figure S9G-L).

Analysis of open reading frame (ORF) types revealed alterations in their distribution. Compared

to w^{1118}, $Mettl5^{1bp}$ exhibited an increased proportion of overlapping downstream ORFs (dORFs) and

fewer non-overlapping dORFs (Figure S10B). In both groups, translated upstream ORFs (uORFs)

were significantly shorter than untranslated uORFs (P < 0.05) (Figure S10C).

Additionally, $Mettl5^{1bp}$ displayed reduced read counts for both dORFs and uORFs (Figure S10D-E).

Based on translation potential, uORFs were classified as either translated or untranslated, and motif

analysis was performed separately for each category (Figure S10F-G).

Mutation of *Mettl5* altered codon preference

Comparison of codon occupancy (A-site) between the two groups revealed

that $Mettl51bp$ preferentially used GAC and GAU, whereas $w1118$ favored UCC (Figure S10H). This

trend was further supported by the cumulative frequency distribution of these codons (Figure S10I-K).

Since GAC and GAU both encode aspartate (Asp), we analyzed Asp amino acid occupancy.

Intriguingly, Asp was significantly enriched in $w1118$ during translation (Figure S10L), suggesting

that $Mettl51bp$ may exhibit altered translational regulation (see Discussion for details).

Metagene analysis of RPFs revealed distinct translation patterns between $Mettl5^{1bp}$ and w^{1118}. The

coding sequence (CDS) and flanking regions were segmented into 100 equal bins, and average RPF

density was computed for each bin. The resulting plots illustrate differences in ribosome occupancy

between $Mettl5^{1bp}$ and w^{1118} along the CDS (Figure S10M), near the translation start site (Figure S10N)

and around the translation termination site (Figure S10O). Notably, $Mettl5^{1bp}$ and w^{1118} exhibited

divergent ribosome occupancy patterns, particularly along the CDS and near the start codon (Figure

S10M, N), suggesting potential differences in translation dynamics and initiation efficiency.

Mettl5 regulates the clock gene regulatory loop

Our findings demonstrate that _Mettl51bp_ disrupts the core clock gene regulatory loop controlling

circadian rhythm. We observed significant alterations in both transcriptional and translational levels

of multiple clock genes, with _Cry_ and _Clock_ showing upregulation while _Tim_, _per_, _vri_, and _pdp1_ were

downregulated at both levels (Figure 4A-B). Notably, _per_, _vri_, and _pdp1_ exhibited particularly

pronounced downregulation in translation efficiency (Figure 4C). These changes occurred without

affecting clock neuron morphology at different time points (Figure S11).

The observed expression patterns revealed an unexpected regulatory relationship: while the

canonical clock circuitry positions _Per_ downstream of _Clock_, our finding that _Clock_ was upregulated

while _per_ was downregulated suggests _Per_ may actually function upstream of _Clock_ in _Mettl5_-

mediated regulation. Surprisingly, despite the transcriptional downregulation of _per_ (Figure 4D), we

detected increased PER protein levels through both immunostaining and western blot analyses (Figure

4E-H). Detection of the PER protein at different time points indicated that it was increased at both ZT

0 and ZT18 (Figure 4E-F). This apparent contradiction aligns with the observed circadian phenotype,

as _Mettl5_ mutants showed significantly longer period lengths (Table 1), mirroring effects seen when

PER stabilization results from reduced kinase activity as previously reported (Philpott et al., 2023).

Genetic epistasis experiments further supported this model, with clock gene mutants modified _Mettl5_

mutant phenotypes that suggesting both _Clock_ and _Per_ downstream of _Mettl5_ (Figure 4I-N, Table 1).

Secondary effect may exist for the significant increase in daytime sleep in the double mutants. Together,

these results indicate that _Mettl5^{1bp}_ affects circadian regulation through mechanisms that extend

beyond transcriptional control, likely involving post-translational regulation of PER protein stability.

To investigate the factors mediating PER protein level changes in _Mettl5_ mutants, we examined

the ubiquitin-proteasome pathway, which plays a well-documented role in Period protein degradation

(Grima et al., 2002; Ko et al., 2002; Chiu et al., 2008). Notably, recent evidence indicates that N6-

methyladenosine (m6A) regulates the ubiquitin-proteasome system in other biological contexts (Sun

et al., 2023). Our integrated analysis of RNA-seq and Ribo-seq data revealed significant

downregulation of multiple proteasome pathway components in _Mettl5^{1bp}_ mutants at both

transcriptional and translational levels (Figures 3E, 3I, 4O), suggesting impaired protein degradation

capacity. Based on these findings, we propose a model where *Mettl5* regulates circadian function

through three interconnected mechanisms: first, by directly modulating proteasome components to

control PER protein stability post-translationally; second, by transcriptionally and translationally

regulating Per and other clock expression. In this model, $Mettl5^{1bp}$-induced proteasome

downregulation leads to PER accumulation, which is responsible for the phenotypes (Figure 5).

$Mettl5^{1bp}$ alters axon complexity

Previous studies have established a strong correlation between sleep homeostasis and synaptic

complexity (Bushey et al., 2011). Additionally, synaptogenesis has been shown to enhance proteasome

activity in axons (Costa et al., 2019). Given our observations of impaired sleep rebound following

deprivation and altered expression of proteasome subunits in $Mettl5^{1bp}$ mutants, we sought to examine

potential effects on synaptic complexity.

To assess synaptic complexity, we adapted an established quantification method (Bushey et al., 2011)

using Syt-GFP, a marker that colocalizes with endogenous synaptic vesicles, to visualize presynaptic

morphology changes. Control experiments with UAS-*Fmr1* and *Fmr1* mutations successfully

replicated the expected decrease and increase in syt-eGFP signal, respectively (Figure 6A-D).

Strikingly, $Mettl5^{1bp}$ mutants exhibited significantly increased syt-eGFP fluorescence in presynaptic

terminals (Figure 6E-M), indicating altered synaptic complexity.

Discussion

Our study reveals that Mettl5, a known rRNA methyltransferase, modulates sleep through its RNA

methylation activity. Through integrated RNA-seq and Ribo-seq analyses of $Mettl5^{1bp}$ mutants, we

identified *Mettl5*'s downstream targets at both transcriptional and translational levels. Further

investigation demonstrated that Mettl5 influences sleep regulation by affecting two key pathways: the

circadian clock gene network and the proteasome system. These findings provide novel mechanistic

insights into sleep control, highlighting the coordinated role of protein synthesis and degradation in

this process. Notably, our Ribo-seq analysis revealed that $Mettl5^{1bp}$ alters fundamental translation

features, including uORF translation efficiency and codon preference, suggesting rRNA methylation

plays a regulatory role in these processes.

This discovery has important clinical implications, as METTL5, the human ortholog of Mettl5, is

associated with intellectual disability (ID) when mutated (Richard et al., 2019). Our work expands the

understanding of Mettl5's molecular function and may inform potential therapeutic strategies for ID.

The clinical relevance of our findings is underscored by reports of sleep disturbances, particularly

reduced sleep duration, in ID patients - a phenotype that parallels our observations in $Mettl5^{1bp}$ mutants.

The mechanistic framework established in this study could explain these clinical sleep abnormalities.

However, further validation in vertebrate models is needed to determine whether this regulatory

mechanism is evolutionarily conserved and applicable to human sleep disorders.

As shown in Table 1, the $Mettl5^{1bp}/+$ mutant exhibits a robust long-period phenotype, with circadian

rhythms significantly extended to 28.3 ± 0.4 hours compared to the wild-type's 23.9 ± 0.05 hours.

This prolonged period perfectly aligns with the observed behavioral phenotypes, including delayed

nighttime sleep onset, later daytime waking, and the overall shift in sleep profile. This is indeed quite

similar to previous report on Period3 variant (*Zhang et al., 2016*). We think that the prolonged

circadian period contributes to the observed sleep phenotype. However, since total sleep time was

significantly reduced in the mutant, we cannot attribute the phenotype solely to period lengthening.

Furthermore, our 24-hour PER expression analysis in *Mettl5* mutants revealed elevated PER protein

levels at ZT1 and ZT18, while ZT6 and ZT12 showed no significant changes, with no apparent phase

shift. These findings collectively suggest that the phenotype primarily results from PER protein

stabilization and accumulation.

We found that *Mettl5* heterozygotes showed significant reductions in total RNA and 18S rRNA

methylation levels, contrasting with mouse studies where heterozygous knockouts maintained normal

18S rRNA m6A methylation (*Sepich-Poore et al., 2022*). This discrepancy may stem from either

fundamental differences in rRNA methylation regulation between *Drosophila* and mice, or distinct

biological consequences of knockdown versus knockout approaches, as complete gene elimination

often triggers compensatory mechanisms (*Teng et al., 2013; Rossi et al., 2015; Vu et al., 2015; Ma et*

al., 2019; El-Brolosy et al., 2019).

Our study uncovers a previously unrecognized connection between circadian clock genes and

proteasome function. While previous work demonstrated that the circadian clock rhythmically

regulates proteasome components in *Drosophila* fat bodies under dietary restriction (*Hwangbo et al.,*

2023), we now show that *Mettl5* modulates clock protein synthesis and degradation in clock neurons

by influencing proteasome activity. This regulation likely occurs through *Mettl5*-mediated ribosomal

methylation in clock neurons, which impacts the proteasome degradation pathway (*Costa et al., 2019*),

ultimately altering clock protein dynamics.

Mettl5 represents a novel integrator of rRNA methylation and proteasome function, providing a

mechanism to balance protein synthesis and degradation. Interestingly, we observed specific effects

on both transcriptional and translational outputs, with particular proteasome subunits showing

differential regulation. This specificity may arise from either selective translational control by Mettl5

or additional layers of regulation through protein-protein interactions. These findings suggest an

intricate regulatory network coordinating these processes, though further studies are needed to

elucidate the underlying mechanisms.

Our results reveal complex relationships between clock genes and sleep regulation. While *cyc* loss-

of-function alleles show enhanced sleep rebound in females (*Shaw et al., 2002*), *Mettl5*[1bp] mutants

with elevated PER protein exhibit reduced rebound. Notably, *Mettl5*'s regulatory effects display tissue

specificity - although absent in canonical clock neurons, *Mettl5*-Gal4 is expressed in distinct neurons

and glia that appear crucial for sleep regulation. This expression pattern, combined with Clock's

broader distribution (*Patop et al., 2023*), suggests specialized circuits for sleep homeostasis that

warrant further investigation.

Evidence indicates that sleep functions in development, metabolism and neuronal plasticity (*Anafi*

et al., 2019). Sleep exerts effects on neuronal plasticity by modifying synapses. The synaptic

homeostasis theory proposes that sleep has a role in downscale synaptic strength (*Tononi and Cirelli,*

2006). Indeed, synapse markers progressively decrease during sleep in both mammals and *Drosophila*

brains (*López-Varea et al., 2021; Vyazovskiy et al., 2008; Gilestro et al., 2009; Liu et al., 2010*) More

importantly, evidence indicates that sleep need and protein levels in synapse are tightly linked. Studies

of the presynaptic active zone have shown that synaptic plasticity regulates sleep homeostasis (*Huang*

et al., 2020). Genome-wide proteomic studies of synapse in mouse brain indicate that the synaptic

proteins peak around dusk and dawn (*Noya et al., 2019; Brüning et al., 2019*). Sleep deprivation

experiments have demonstrated that the sleep drive is significantly dependent on the cycling of

proteins and phosphoproteins in synapses, in contrast to mRNAs (*Noya et al., 2019; Brüning et al.,*

2019). These results suggest that the gene expression regulation at the protein level is crucial for sleep.

Consistently, we detected changes in the quantity of axons in *Mettl5*^{1bp}, suggesting a possible

regulation of neural circuits by *Mettl5*.

The mechanism by which METTL5 regulates translation warrants further investigation. Previous

studies have demonstrated that METTL5 influences translation (Rong et al., 2020; Peng et al., 2022),

but whether the mechanisms identified here are conserved across other systems remains an intriguing

question. In our analysis, we observed increased usage of aspartate (Asp) codons in *Mettl5* mutants.

Notably, prior work has linked codon usage to PER protein function—specifically, a codon-optimized

version of PER failed to rescue circadian rhythmicity in *per* mutant flies, unlike the wild-type version

(Fu et al., 2016). Further analysis revealed that PER protein levels were elevated in these mutants,

suggesting that codon optimization enhances PER expression (Figure 2B in Fu et al., 2016). Strikingly,

when we examined the codon-optimized region from Fu et al. (2016), we found that GAC (Asp) was

highly enriched, raising the possibility that *Mettl5* mutation affects PER protein accumulation by

altering GAC codon usage. Additional experiments will be needed to validate this hypothesis.

Furthermore, we detected changes in upstream open reading frames (uORFs) in *Mettl5* mutants, but

their relationship to translational regulation requires further exploration.

Our study demonstrates that ribosomal components can exert gene-specific regulatory functions,

building upon previous work showing context-dependent ribosome specialization (*Simsek et al., 2017*).

In *Mettl5* mutants, we observed distinct alterations in the translational efficiency of specific genes.

These effects could potentially arise through two non-exclusive mechanisms: (1) *Mettl5*-mediated

rRNA modifications may modulate ribosomal binding affinity for particular mRNA sequences, or (2)

these modifications might contribute to the formation of specialized ribosome populations that

preferentially translate specific subsets of mRNAs. Future studies will be required to distinguish

between these possibilities and fully elucidate the underlying molecular mechanisms.

Material and methods

Drosophila **Strains**

Fly stocks used in this study were maintained under standard culture conditions. We used the w^{1118} as

the control strain. The following flies were obtained from Bloomington Stock Center: w^{1118} (Bl: 5905),

Mettl5-Gal4 (Bl: 19514), *Repo*-Gal4 (Bl: 7415), *nSyb*-Gal4 (Bl: 51941), *pdf*-Gal4 (Bl: 41286), Clk^{JRK}

(Bl: 80927), per^{01}(Bl: 80917), $Fmr1^{\Delta 50M}$(Bl: 6930)，UAS-*Trmt112*RNAi (VDRC: 101515), UAS-

DenMark,UAS-*syt*.eGFP (Bl: 33064), UAS-GFPstinger (from Yi Rao's lab), UAS-*Fmr1* (Bl: 6931),

UAS-*Mettl5* (FlyORF: F000760) are from Fly ORF collection. UAS-*Mettl5m*-3HA (generated in this

study).

$Mettl5^{1bp}$ and $Mettl5^{9bp}$ mutants were generated by the CRISPR/Cas system as described

previously (*Cheng et al., 2020*). The target gRNA was designed with an online tool:

http://tools.flycrispr.molbio.wisc.edu/targetFinder/. A target sequence was chosen that has the

sequence of 5'- GAGTGGGGTACTGTTCCAAC**AGG** with the PAM sequence in bold. Mutations

were verified by genomic PCR and sequencing.

Sleep behavior assays

All sleep assays were conducted in a controlled environment incubator maintained at 25 ± 1°C

with 60% ± 5% relative humidity. We maintained a 12:12 light-dark cycle with lights on at ZT0 (06:30)

and off at ZT12 (18:30). Fly activity was monitored using the Drosophila Activity Monitoring System

(Trikinetics, Waltham, MA). Following a 2-day acclimation period, we recorded locomotor activity at

1-minute intervals for 3 consecutive days. Data analysis was performed using pysolo (*Gilestro and*

Cirelli, 2009), with sleep defined as ≥5 minutes of continuous inactivity.

Mechanical sleep deprivation was performed using the SNAP method to keep flies awake for 12

h overnight (*Shaw et al., 2002*). Sleep deprivation is applied by mechanical stimuli using a timer-

controlled rotating shaker. The intensity of stimuli was 1500 rpm once every minute (2 s/1 min)

(*Shimizu et al., 2008*). Baseline sleep was established from 24-hour recordings prior to deprivation.

Sleep loss and recovery were quantified according to previous publication (Cirelli C et al., 2005).

Regarding the 'awake %' metric, it indicates that at specific time points (e.g., ZT14), the

percentage of awake fruit fly population at that moment. At ZT19, we evaluated arousal thresholds by

administering a standardized gentle mechanical stimulus. Responsiveness was determined by

monitoring activity for 1 minutes post-stimulation. Flies showing no activity during this window were

scored as non-responsive. We calculated arousal percentages from the proportion of flies that awakened in response to stimulation.

Circadian rhythm of individual male flies was measured using the Drosophila Activity Monitoring (DAM) System (Trikinetics). Male flies were loaded individually into glass tubes with a length of 65 mm and an inner diameter of 5 mm. The tubes contained standard cornmeal fly food at one end and were sealed with a cotton stopper at the other end. The flies were entrained to a 12 h light/12 h dark cycle for 3 days and then released to constant darkness for at least six days to measure their rhythmicity.Data analysis is done on a Macintosh computer running the FaasX (Fly activity analysis suite) software.

Statistical analysis

All statistical analyses were performed using GraphPad Prism 5 software. For sleep parameter comparisons, we used nonparametric tests including the two-tailed Mann-Whitney test for pairwise comparisons and one-way ANOVA with Tukey's post hoc test for multiple comparisons. qPCR data were analyzed using unpaired Student's t-tests, while lifespan data were evaluated with Log-rank (Mantel-Cox) tests. The specific statistical test used for each experiment is indicated in the corresponding figure. In all analyses, a p-value of less than 0.05 was considered statistically significant.

Quantitative PCR

Total RNA was extracted from cells and tissues using the TRNzol Universal Reagent (Tiangen #DP4-02). For cDNA synthesis, we employed the PrimeScript RT reagent Kit with gDNA Eraser (TAKARA #RR047A) following the manufacturer's protocol. Quantitative PCR was carried out using SuperReal PreMix Plus (SYBR Green) (Tiangen #DP4-02). *RP49* served as the endogenous reference gene for normalization across samples. All experiments included three independent biological replicates to ensure reproducibility. The primers used in this experiment are *RP49*-F: CGGTTACGGATCGAACAAGC; *RP49*-R: CTTGCGCTTCTTGGAGGAGA; *Mettl5*-F: CGGTTTCTGGAGGTGGC; *Mettl5*-R: GCTGGCGTCGATGTTGTAC..

Imaging and analysis of axon volume

We quantified small LNvs axon volume using ImageJ's Object Counter 3D plugin to measure pixel counts from raw images. Image processing involved applying a standardized threshold that clearly

visualized intact axons in control samples. For consistent measurements, we specifically analyzed the

axon span between the first axonal bifurcation and the terminal tip, as indicated by the white

rectangular markers in Figure 8.

Sample collection and library construction for Ribo-seq and RNAseq

For sample collection, w^{1118} and *Mettl5^{1bp} Drosophila* were harvested at ZT15, immediately

transferred to centrifuge tubes, and flash-frozen in liquid nitrogen. Fly heads were subsequently

separated and collected while frozen, with three biological replicates prepared for each genotype. Each

sample was equally divided for parallel RNA-seq and Ribo-seq analyses. For Ribo-seq library

preparation, samples were lysed in buffer containing 50 mg/mL cycloheximide (Novogen, China) to

preserve ribosome positioning, followed by RNase I digestion to generate ribosome-protected

fragments (RPFs). Monosomes were isolated using MicroSpin S-400 HR size-exclusion

chromatography, followed by rRNA depletion and PAGE purification to select 20-38 nt RPFs. Purified

fragments underwent end repair, adapter ligation, reverse transcription, and PCR amplification before

Illumina PE150 sequencing. For RNA-seq, total RNA was extracted using TRIzol reagent, with cDNA

libraries prepared and sequenced using Illumina PE150 by Novogen. Raw sequencing data in FASTQ

format were processed to remove adapter sequences, reads containing N bases, and low-quality reads,

while simultaneously calculating Q20/Q30 scores and GC content to generate clean reads for

downstream analysis.

RNA-seq analysis

We performed genome alignment and transcriptome analysis using the following pipeline: First,

we built a Hisat2 index (v2.0.5) for the Drosophila melanogaster reference genome (dm6 assembly).

Clean paired-end reads were then aligned to this reference using Hisat2 (v2.0.5) (*Kim et al., 2019*).

The resulting alignments were processed using StringTie (v1.3.3b) (*Pertea et al., 2015*) for reference-

based transcript assembly. For gene-level quantification, we used FeatureCounts (v1.5.0) (*Liao et al.,*

2014) to count reads mapped to each annotated gene. Differential expression analysis between w^{1118}

and *Mettl5^{1bp}* was conducted with DESeq2 (v1.20.0) (*Pertea et al., 2015*), with genes meeting both

criteria (adjusted p-value < 0.05 and absolute log2 fold change ≥ 1) considered statistically significant.

Ribo-seq analysis

We implemented a comprehensive Ribo-seq analysis pipeline beginning with quality filtering using Bowtie (*Liao et al., 2014*) to remove reads aligning to non-coding RNAs (rRNA, tRNA, snoRNA, snRNA from FlyBase Release 6.13) with a 2-mismatch allowance (-v 2). The remaining ribosome-protected fragments (RPFs) were mapped to the *Drosophila melanogaster* genome (FlyBase Release 6.13) using STAR (v2.7.3a) (Dobin et al., 2013), followed by transcript-level alignment to protein-coding sequences using Bowtie (v1.2.2) (*Langmead et al., 2009*) with parameters "-a -v 2". CDS-aligned RPFs were quantified using featureCounts (Subread v1.6.3) and normalized as RPKM. Differential expression analysis was performed with DESeq2 (v1.14.1) (*Love et al., 2014*) using thresholds of $|log2FC| \geq 0.265$ and padj < 0.05, while translation efficiency differences were assessed using RiboDiff. Data quality was verified through riboWaltz (v1.1.0) (*Lauria et al., 2018*) for 3-nt periodicity and reading frame analysis, with Ribocode (*Xiao et al., 2018*) employed for P-site positioning and uORF motif analysis. Functional enrichment analyses (GO, KEGG, GSEA) were conducted using clusterProfiler (v4.5.2.002) (Wu et al., 2021), with results visualized through ggplot2-generated plots and GO term simplification performed using SimplifyEnrichment (*Gu and Hübschmann, 2022*).

Immunofluorescence experiments

We performed immunofluorescence staining on 7-15 days old adult flies (unless otherwise specified). Flies were anesthetized with CO_2 and dissected in ice-cold 0.03% PBST (1× PBS with 0.03% Triton X-100; Sigma, T9284). Samples were fixed in 2% paraformaldehyde (PFA) for 55 minutes at room temperature (RT), followed by four 15-minute washes in 0.03% PBST at RT. After blocking overnight at 4°C in 10% Normal Goat Serum (NGS; in 1× PBS with 2% Triton), samples were incubated with primary antibodies for 24 hours at 4°C. Primary antibodies included rat anti-Elav (DSHB, 9F8A9; 1:200) and mouse anti-Repo (DSHB, 8D12; 1:200), diluted in antibody buffer (1.25% PBST, 1% NGS). Following four 15-minute washes in 1× PBS with 1% Triton at RT, samples were incubated overnight at 4°C with secondary antibodies: Alexa Fluor™ 568 (Thermo Fisher, A11004; 1:200) and Alexa Fluor™ 647 (Thermo Fisher, A21247; 1:200). After four additional 15-minute washes in 1× PBS with 1% Triton at RT, samples were mounted using DAPI-containing antifade mounting medium (Solarbio, S2110).

Images were acquired using a Leica SP8 confocal microscope with LAS X software, applying auto Z-

brightness correction when needed for signal uniformity. Images were processed in Adobe Photoshop

CS6 and figures assembled in Adobe Illustrator 2020. PDF immunofluorescence followed the same

protocol. All experiments included ≥3 biological replicates, each containing ≥10 flies. Fluorescence

intensity was quantified using ImageJ.

Western Blotting

Protein samples were prepared by homogenizing approximately 30 fly heads in RIPA lysis buffer

(150 mM NaCl, 1.0% NP-40, 0.5% sodium deoxycholate, 0.1% SDS, 50 mM Tris-HCl, pH 8.0)

supplemented with protease (CW2200S) and phosphatase (CW2383S) inhibitor cocktails according

to manufacturer specifications. Lysates were mixed with 2× SDS loading buffer, boiled at 100°C for

5 min, and immediately cooled on ice.

For immunoblotting, membranes were probed with rabbit anti-PER primary antibody (1:5000

dilution; kindly provided by Dr. Jeffrey Price's laboratory, University of Missouri-Kansas City)

overnight at 4°C, followed by incubation with HRP-conjugated goat anti-rabbit IgG secondary

antibody (1:1500; ABclonal, AS014) for 4 hr at room temperature. Protein signals were detected using

ECL substrate (ABclonal, RM00021P) and imaged with an Amersham ImageQuant 800 system (GE

Healthcare). Band intensities were quantified using ImageJ software, with three biological replicates

performed for statistical reliability.

LC-MS/MS analysis of m^{6}A levels

Total RNA was isolated from cells and tissues using TRNzol Universal Reagent (Tiangen, #DP4-02).

For 18S rRNA purification, we separated total RNA by polyacrylamide gel electrophoresis followed

by gel extraction. Our analysis utilized ribonucleoside standards including adenosine (rA) and N6-

methyladenosine (N6 mA), with a mobile phase consisting of methanol:ddH2O (v/v).

For each biological replicate, 1 μg of total RNA or purified 18S rRNA was digested to single

nucleosides using Nucleoside Digestion Mix (NEB, #M0649). Proteins were precipitated by adding a

4:1 methanol:digest ratio and incubating at -20°C for 2 hours. Quantitative analysis was performed

using multiple reaction monitoring (MRM) with the following transitions: 268.10275→136.0621 (rA),

282.11835→150.0774 (N6 mA).

Negative geotaxis RING assay

Climbing assays were performed following a 12-hour recovery period after CO_2 anesthesia, using 200 flies per genotype distributed across 10 vials (20 flies/vial) marked at 90 mm height. Each assay consisted of three trials separated by 15-minute intervals, initiated by sharply tapping vials three times to induce negative geotaxis, with the number of flies reaching the 90 mm mark within 10 seconds recorded per trial. Five groups were tested per genotype (15 trials total). All assays were video recorded under standardized conditions with the camera positioned 20 cm from vials and uniform backlighting provided by a white open-faced box to ensure consistent imaging quality.

Data availability

The RNA-seq and Ribo-seq data from this publication have been deposited to the NCBI bioproject database https://www.ncbi.nlm.nih.gov/bioproject/ and assigned the identifier PRJNA994860.

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Author contributions

XY, XW and TF designed and performed the experiments, analyzed the data and wrote the manuscript. XY performed the bioinformatics analysis, analyzed the data and edited the manuscript. YR analyzed data and edited the manuscript. JD designed experiments, analyzed data, wrote and edited the manuscript. All authors have read and approved the final manuscript.

Declaration of interest

The authors declare that they have no conflict of interest.

Supplementary information

Supplementary Figures S1-S11

Figure legends

Figure 1 *Mettl5* is a regulator of *Drosophila* sleep.

(A) Diagram illustrating CRISPR-Cas knockout of 1 or 9 bases in the *Mettl5* gene. The corresponding protein sequence is listed with the predicted N6 adenine specific nucleic acids methyltransferase domain highlighted in the red box. (B) Relative expression of *Mettl5* mRNA in homozygous *Mettl5^1bp^* and *Mettl5^9bp^* mutant male flies compared to control flies. (C) Sleep curve throughout the day for *Mettl5* mutant male flies (blue) and control flies (black). (D) Total sleep of *Mettl5* mutant male flies and control flies in 24 hours. (E) Total sleep of *Mettl5* mutant male flies and control flies within day and night, respectively. (F) Sleep bout duration of *Mettl5* mutant male flies and control flies. (G) Number of sleep bouts of *Mettl5* mutant male flies and control flies. (H) Percentage of awake for *Mettl5* mutant flies and control flies. (I) Sleep curve is tracked throughout the entire day prior to sleep deprivation and during the daytime sleep rebound period. (J) *Mettl5* mRNA expression level at different time points. W(wake), SD (sleep deprivation), SR (sleep recovery). (K) Sleep curve is tracked throughout the entire day prior to sleep deprivation and during the daytime sleep rebound period in *Mettl5* mutant male flies (blue) and control flies (black). (L) Response to sleep deprivation and performance measures in *Mettl5* mutants and controls. Black bars represent the amount of sleep lost during the 24-hour sleep deprivation period, blue bars indicate the amount of sleep regained, whereas the red bars indicate the proportion of sleep recovered (right y-axis). (M) Sleep arousal of *Mettl5^1bp^* male flies and control flies at ZT19. (N) Sleep curve throughout the day for the following genotypes: w^{1118}(black), *Mettl5^1bp^/+* (blue)*, and Mettl5-Gal4, UAS-Mettl5, Mettl5^1bp^/+*(pink). (O) Total sleep of the indicated genotypes. (P) Sleep bout duration of the indicated genotypes. (Q) Number of sleep bouts of the indicated genotypes. (R) Percentage of awake for the indicated genotypes. For * stands for p<0.05, ** stands for p<0.01, *** stands for p<0.001, *ns* stands for not significant. For letter-based annotations, Groups with no significant differences share the same letter, Groups with significant differences are assigned new letters.

Figure 2 *Mettl5* regulation of *Drosophila* sleep was dependent on its methyltransferase activity.

(A) The m⁶A level in the total RNA of *Mettl5* mutant male flies. (B) The m⁶A level in the 18S rRNA of *Mettl5* mutant male flies. (C) The sleep curve throughout the day shows the sleep pattern of induced

Trmt112 RNAi male flies and control flies. (D) Total sleep of induced *Trmt112* RNAi male flies and

control flies. (E) Sleep bout duration in induced *Trmt112* RNAi male flies and control flies. (F)

Number of sleep bouts in induced *Trmt112* RNAi male flies and control flies. (G) Percentage of awake

in *Trmt112* RNAi and control flies. (H) Sleep curve throughout the day for *Mettl5* mutant male flies,

induced *Mettl5m* over expression male flies and control flies. (I) Total sleep of *Mettl5* mutant male

flies, induced *Mettl5m* over expression male flies and control flies. (J) Sleep bout duration in *Mettl5*

mutant male flies, induced *Mettl5* over expression male flies and control flies. (K) Number of sleep

bouts in *Mettl5* mutant male flies, induced *Mettl5* over expression male flies and control flies. (L)

Percentage of awake in *Mettl5* mutant male flies, induced *Mettl5* over expression male flies and control

flies. For * stands for p<0.05, ** stands for p<0.01, *** stands for p<0.001, *ns* stands for not significant.

For letter-based annotations, Groups with no significant differences share the same letter, Groups with

significant differences are assigned new letters.

Figure 3 RNA-seq and Ribo-seq analysis revealed changes in the gene profile of *Mettl5^{1bp}*.

(A) Venn diagram depicting the number of significant differentially expressed genes revealed by RNA-

seq and Ribo-seq. (B) Volcano plot representing the differentially expressed genes identified by RNA-

seq. Genes that met the criteria of |log2(fold-change)| ≥ 1 and p.adjust < 0.05 were considered

significantly expressed, marked in orange for downregulation and green for upregulation, comparing

with the controls. (C) Volcano plot representing the differentially expressed genes identified by Ribo-

seq. Candidates that satisfied the criteria of |log $_2$ (fold-change) | ≥ 0.265 and p.adjust <0.05 were

regarded as significantly expressed, marked in red for downregulation and blue for upregulation,

respectively. (D-K) Gene set enrichment analysis of differentially expressed genes revealed by RNA-

seq (D-G) and Ribo-seq (H-K). All the plots are generated using the KEGG gene set database. The bar

chart at the bottom of each panel shows the distribution of target genes for each pathway according to

their rank position. Each vertical line represents a gene. Genes on the left show positive

correlation with *Mettl5^{1bp}*, while, genes on the right show negative correlation with *Mettl5^{1bp}*. The

green line indicates the enrichment score (ES), and NES stands for normalized enrichment score. (L)

Distribution of the differentially expressed genes revealed by both RNA-seq and Ribo-seq. (M)

Cumulative distribution of TE frequencies among *w^{1118}* and *Mettl5^{1bp}* (N, O) GO and KEGG

enrichment of significantly changed TE-related genes between w^{1118} and $Mettl5^{1bp}$. The color of the

bar indicates the enrichment p.adjust value. (P) KEGG network showing the top 10 pathway and

associated genes. The size of the dots represents the number of genes in the pathway.

Figure 4. Clock genes expression mediated the sleep phenotype caused by *Mettl5* mutation.

(A-C) Fold changes in clock genes with significant expression level differences between w^{1118} and

$Mettl5^{1bp}$ were observed in RNA-seq, Ribo-seq, and translation efficiency analyses. (D) The gene

expression levels of per at 4 different timepoints of w^{1118} and $Mettl5^{1bp}$. (E) Western blot analysis of

PER protein and Tubulin protein at 4 different timepoints, (F) along with their ratio quantification. (G)

Representative image of PER protein immunofluorescence staining at ZT0 in the small ventral lateral

neurons (small LNvs). (H) Statistical analysis of the immunofluorescence intensity for PER in small

LNvs. (I) Sleep curve throughout the day for $Mettl5^{1bp}$, per^{01}, double mutant and control flies. (J)

Total sleep for $Mettl5^{1bp}$, per^{01}, double mutant and control flies. (K) Percentage of awake time in

$Mettl5^{1bp}$ flies, partially rescued by double mutant flies. (L) Sleep curve throughout the day for

$Mettl5^{1bp}$, per^{01}, double mutant and control flies. (M) Total sleep for $Mettl5^{1bp}$, per^{01}, double mutant

and control flies. (N) Percentage of awake time in $Mettl5^{1bp}$ flies, partially rescued by double mutant

flies. (O) Fold changes in proteasome subunits with significant expression level differences between

w^{1118} and $Mettl5^{1bp}$ were observed in RNA-seq, Ribo-seq, and translation efficiency analyses.

For statistical significance, * stands for p<0.05, ** stands for p<0.01, *** stands for p<0.001, *ns* stands

for not significant.

Figure 5 A working model illustrating the role of *Mettl5* in *Drosophila* sleep was presented.

Figure 6 The axon complexity was found to be affected by *Mettl5^{1bp}*

(A–C) Representative confocal micrographs of small ventral lateral neuron (s-LNv) axonal terminals

across different genotypes. Presynaptic structures were visualized using *syt*-eGFP (green), which

colocalizes with endogenous synaptic vesicles: (A) control (B) Fmr1 overexpression and (C) Fmr1

null mutant. (D) Quantification of s-LNv axonal terminal volumes corresponding to genotypes in (A–

C). (E–H) Representative images of axonal terminal morphology in control flies at four time points:

ZT0, ZT6, ZT12, and ZT18. (I–L) Representative images of axonal terminal morphology in *Mettl5*

mutant flies at ZT0, ZT6, ZT12, and ZT18. (M) Quantitative comparison of axonal terminal volumes

between control and Mettl5 mutant flies at different time points. Scale bar: 50 μm. For statistical

significance, * stands for p<0.05, ** stands for p<0.01, *** stands for p<0.001, *ns* stands for not

significant.

Table 1 Circadian rhythm phenotypes of various mutants.

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Mettl5 allele verification

Sleep phenotypes

Figure 3

Differential expression of clock genes

Differential expression of protesome genes

RNA and protein expression of per

Genetic interactions

Genotype	NumTotal	%Rhythmic	Period	Power
w^{1118}	32	92.6	23.9± 0.05	127.9± 7.62
$Mettl5^{1bp}$/+	32	92.3	28.3± 0.4 ***	114.3± 6.76 ns
$Mettl5^{1bp}$/+ ; UAS-$Mettl5$/$Mettl5$Gal4	32	96.9	24± 0.02 ns	127.1± 5.42 ns
$Mettl5^{1bp}$/Clk^{JRK}	44	29.5	24± 0.04 ns	68.5± 8.85 ***
$Mettl5^{1bp}$/Per^{01}	54	9.3	24.2± 0.2 ns	43.8± 13.41 ***

Figure S4

Figure S7

Figure S8

Figure S9

Figure S10

Supplementary figure legends

Figure S1 *Mettl5* functions in a portion of neurons and glia.

(A-C) Colocalization of Mettl5-YFP with antibodies staining ELAV. Scale bar: 50 μm (main figures) and 20 μm (zoomed regions) (D-F) Colocalization of Mettl5-YFP with antibodies staining REPO. Scale bar: 50 μm (main figures) and 20 μm (zoomed regions). (G) Sleep latency. (H) Climbing index of *Mettl5^{1bp}* male flies and control flies. (I) Sleep curves throughout the day for *Mettl5*-RNAi and control flies. (J) Total sleep for *Mettl5*-RNAi and control flies flies. (K) Sleep bout duration of *Mettl5*-RNAi and control flies. (L)Number of sleep bouts of *Mettl5*-RNAi and control flies. For statistical significance, * stands for p<0.05, ** stands for p<0.01, *** stands for p<0.001, *ns* stands for not significant

Figure S2 Quality control of RNA-seq and Ribo-seq samples.

(A, D) Meta analysis of gene expression data obtained from RNA-seq and Ribo-seq. (B, E) Principal component analysis (PCA) of different samples for RNA-seq and Ribo-seq. (C, F) Correlation heatmap of normalized RNA and RPFs values between every pair of samples. The color and size of dots represent the Pearson correlation coefficient.

Figure S3 Heatmap of differentially expressed genes (DEGs) between *Mettl5^{1bp}* and *w^{1118}*.

Heatmap showing the top 100 significantly changed genes identified by RNA-seq (A) and Ribo-seq (B) in *Mettl5^{1bp}*. Red indicates increased expression, while blue indicates decreased expression.

Figure S4 Enrichment analysis of differential expressed genes between *Mettl5^{1bp}* and *w^{1118}*.

(A-D) GO analysis performed on the differentially expressed genes revealed by RNA-seq (A, C) and Ribo-seq (B, D). The bubble chart shows the top 15 most significantly enriched categories under the themes of biological processes, cellular components and molecular functions. The circle color indicates the enrichment p.adjust, and the dot size indicates the number of differentially expressed genes in the functional class or pathway.

Figure S5 Simplified GO enrichment analysis for differentially expressed genes in RNA-seq and Ribo-seq.

GO enrichments for biological process (A, D), cellular components (B, E), and molecular function (C, F) were shown. The bottom right clusters, which lack word cloud annotations, include all other small clusters with fewer than 5 terms. The color shades represent the similarity

of GO terms based on binary cut clustering.

Figure S6 Gene set enrichment analysis (GSEA) for differentially expressed genes found in RNA-seq and Ribo-seq.

GSEA were performed on the differentially expressed genes identified in RNA-seq (A, B) and Ribo-seq (C, D). The top 5 terms or pathways found in GSEA based on GO (A, C) and KEGG (B, D) databases are shown. Genes were sorted according to their signed NES values. The circle color indicates the enrichment p.adjust, and the dot size indicates the number of differentially expressed genes in the respective functional class or pathway.

Figure S7 GO and KEGG enrichment analysis of genes that were significantly changed at both transcriptional and translational levels (showed in figure3L).

(A, B) Top 10 GO and KEGG terms for genes with the same direction of change at transcriptional and translational levels. (C, D) Top 10 GO and KEGG terms for genes with the opposite direction of change at transcriptional and translational levels. (E, F) Top 10 GO terms for genes that were only changed at the transcriptional level; (G, H) Top 10 GO terms for genes that were only changed at the translational level. The circle color indicates the enrichment p.adjust, and the dot size indicates the number of differentially expressed genes in the respective functional class or pathway. Changes with $|\log_2$ (fold change) $| \geq 0.265$ of translational levels and $|\log_2($fold change$)| \geq 1$ of transcriptional levels were considered significant.

Figure S8 TE correlation and P-site offsets

(A-F) TE correlation between the three biological replicates of two group showing high reproducibility. Different colors represent the density distribution of different translated genes. (G-L) Metagene analysis of individual 28nt reads mapped to their 5' ends to determine P-site offsets. The distribution of 28nt RPF reads around the ribosomal P site in w^{1118} (G-I) and *Mettl5* (J-L). The horizontal axis represents the distance from the start-codon or stop-codon, and the vertical axis represents the read counts. The P-site positions are colored according to the reading frame.

Figure S9 Global scan of RPFs distribution

(A-F) Distribution of RPFs periodicity among translation start codons and translation end codons in w^{1118} (A-C) and *Mettl5^{1bp}* (D-F). (G-L) Distribution of RPFs on different coding frames, including frame 0, frame 1 and frame 2, for the 5'UTR, CDS and 3'UTR in w^{1118} (G-I)

and *Mettl5^{1bp}* (J-L). The plot illustrates the distribution of ribosome-protected fragments (RPFs) across different coding frames (frame 0, frame 1, and frame 2) on the x-axis, with the y-axis representing the percentage of RPFs in each coding frame.

Figure S10 Ribo-seq has revealed that the *Mettl5^{1bp}* leads to changes in global translation features.

(A) The distribution of RPF length among different samples in *Mettl5^{1bp}* and *w¹¹¹⁸* was compared. (B) A comparison was made between *Mettl5^{1bp}* and *w¹¹¹⁸* regarding the statistics on active translated ORFs. The distribution of different types of ORFs based on their relative location to associated coding sequences (CDS) was illustrated in the comparison between *Mettl5^{1bp}* and *w¹¹¹⁸*. The categories include "annotated" (ORFs that overlap with annotated CDS and have the same stop codon as the annotated CDS), "uORF" (ORFs located upstream of annotated CDS , not overlapping annotated CDS), "dORF" (ORFs located downstream of annotated CDS, not overlapping annotated CDS), "Overlap_uORF" (ORFs located upstream of annotated CDS, overlapping annotated CDS), "Overlap_dORF" (ORFs located downstream of annotated CDS, overlapping annotated CDS), "Internal" (ORFs located within annotated CDS but in a different frame relative to the annotated CDS), and "novel" (ORFs located in non-coding genes or non-coding transcripts of coding genes). The percentages of each ORF category were compared between *Mettl5^{1bp}* and *w¹¹¹⁸*, providing insights into the differences in ORF distribution between the two analyzed samples. (C) A comparison was made between *Mettl5^{1bp}* and *w¹¹¹⁸* regarding the length of translated and untranslated uORFs. (D) The count of dORFs was compared between *Mettl5^{1bp}* and *w¹¹¹⁸*. (E) The count of translated uORFs was compared between *Mettl5^{1bp}* and *w¹¹¹⁸*. (F, G) Motif analysis was performed separately on the translated and untranslated uORFs between *w¹¹¹⁸* and *Mettl5^{1bp}*. (H) The correlation of codon occupancy (A-site) between *Mettl5^{1bp}* and *w¹¹¹⁸* was analyzed. (I-L) The cumulative changes in RPFs for codon UCC (I), GAC (J), GAU (K) and amino acid ASP (L) were compared between *Mettl5^{1bp}* and *w¹¹¹⁸*. The statistical analysis was performed using the Kolmogorov-Smirnov test (KS test). (M-O) A metagene plot was generated for RPFs on whole CDS region and around the CDS start and end regions in group case study. The CDS was divided into 100 equal bins.

Figure S11 Morphology of clock neurons in *Mettl5* mutant.

Ventral lateral neurons labeled by PDF staining in *Mettl5^{1bp}* and control flies across different

time points.