METTL5 Blockade Enhances Anti-Tumor Immune Response via Inducing Neoantigen Generation

Yangyi Zhang[1, #], Xiaoyan Shi[1, #], Yuci Wang[1, 2, #], Ruiqi Wang[1], Yanlan Cao[1], Hao Chen[1, *]

[1]Department of Human Cell Biology and Genetics, SUSTech Homeostatic Medicine Institute, School of Medicine, Southern University of Science and Technology, Shenzhen 518055, Guangdong, China

[2]Peking university, Beijing, 100000, China

[#]These authors contributed equally to this work.

*Corresponding author. chenh7@sustech.edu.cn (H.C.).

ABSTRACT

Tumor neoantigens play a pivotal role in immunotherapy through eliciting tumor-specific immune responses. However, their clinical application is limited by the restricted repertoire of naturally occurring neoantigens, which undergoes continuous immune editing within the tumor microenvironment. To address this critical challenge, we explored translational regulation as a reversible and safer strategy to augment neoantigen production. Through systematic analysis of translational regulatory factors associated with immunotherapy response, we identified METTL5 as a key modulator of anti-tumor immunity. METTL5 depletion in syngeneic tumor models demonstrated significant enhancement of endogenous anti-tumor immunity across distinct MHC-I haplotypes, with synergistic effects observed when combined with immune checkpoint blockade. Based on evolutionary conservation of ribosomal decoding center modifications and established mechanisms from prokaryotic studies, we propose that METTL5-mediated m^6A modification at the small ribosomal subunit decoding center may contribute to translational fidelity regulation. Mechanistically, integrated MHC-I immunoprecipitation-mass spectrometry and ribosome profiling revealed the identification of non-canonical peptides, including those derived from novel unannotated open reading frames (nuORFs), which provide previously absent neoantigen sources for tumor cells. T-cell receptor sequencing further demonstrated significantly enhanced clonal diversity and tumor-specific reactivity of infiltrating T cells in METTL5-deficient tumors. Therapeutic vaccination strategies targeting METTL5-deficiency-induced neoantigens, including both peptide-based and mRNA vaccines, exhibited partial anti-tumor efficacy in murine tumor models. This study establishes METTL5 as a critical translational regulator that restricts neoantigen diversity and proposes a novel combinatorial strategy to enhance immunotherapy through translational control.

INTRODUCTION

The complexity of cancer therapy originates from the high heterogeneity of tumors, dynamic immune escape mechanisms, and the multi-faceted challenges posed by the immunosuppressive tumor microenvironment[1-8]. Conventional therapies (e.g., chemotherapy and radiotherapy) often cause severe toxic side effects with limited therapeutic efficacy, whereas immunotherapy, despite its advantages of low toxicity and durable responses, remains constrained by low response rates and the development of acquired resistance. [3-5,9-16]. Strategies targeting universal or single antigens are particularly vulnerable to failure due to antigen loss or exhaustion, and they fail to address tumor heterogeneity while introducing risks of immune-related adverse events (irAEs)[17-20].

A promising solution lies in the identification of tumor-specific mutations to generate neoantigens, which can activate polyclonal T-cell responses, overcome immune tolerance, and avoid off-target toxicity[21-28]. This "dynamic targeting" approach offers a means to address heterogeneity and drug resistance[4,5,29,30]. Neoantigens, derived from short peptides (8–25 amino acids) resulting from errors in gene expression regulation, are presented on tumor cell surfaces via human leukocyte antigen (HLA) molecules, enabling T-cell recognition[31-34]. Unlike self-antigens, neoantigens are not subject to central tolerance mechanisms and exhibit robust immunogenicity, making them ideal targets for immunotherapy[21,22,25,27,35-37].

Tumor mutational burden (TMB) has been proposed as a key biomarker in cancer immunotherapy, particularly for PD-1/PD-L1 inhibitors[38-42]. Moreover, TMB as a potential biomarker, exhibits cancer-type-dependent predictive value in PD-1/PD-L1 inhibitor treatments (e.g., non-small cell lung cancer and melanoma)[43,44]. Importantly, TMB represents only one pathway contributing to neoantigen generation. Certain tumor mutations enhance the quality and quantity of functional neoantigens, such as those

targeting highly immunogenic mutations like KRAS G12D, IDH1 R132H, and TP53

[45-48]. These neoantigen vaccines have demonstrated the ability to activate T-cell

responses and penetrate the blood-brain barrier in solid tumors such as pancreatic

cancer and glioma. Furthermore, the combination of personalized vaccines (e.g.,

NeoVax) with immune checkpoint inhibitors has advanced clinical exploration in

precision immunotherapy, suggesting that the quality of neoantigens generated by

tumor mutations may guide therapeutic efficacy[49].

However, clinical application of neoantigen-targeted therapies faces multiple

challenges. First, most antigens arising from tumor mutations are of low quality (e.g.,

synonymous mutations or cryptic epitopes), exhibiting insufficient immunogenicity[50-

54]. Second, certain mutated peptides share high sequence homology with self-proteins,

leading to weakened anti-tumor immune responses due to central tolerance (thymic

selection) or peripheral tolerance (T cell exhaustion) mechanisms[16,55]. More critically,

the process of immune editing within the tumor microenvironment continuously

eliminates highly immunogenic epitopes, leaving residual antigen pools dominated by

low-affinity variants that further impair T cell activation efficiency[52,56]. To overcome

these limitations, current intervention strategies-such as CRISPR/Cas9-induced

mutations or DNA-damaging agents to enhance neoantigen load—can transiently

expand neoantigen numbers[5,57,58]. However, their irreversible genomic modifications

risk exacerbating tumor heterogeneity or causing normal tissue toxicity, significantly

hindering clinical translation[59,60].

Compared to genomic interventions, translational regulation offers unique advantages

for neoantigen development: First, the reversibility of translation allows dynamic

control of ribosomal function (e.g., modulating methylation levels at the decoding

center) to flexibly adjust translational error rates, avoiding permanent genomic

alterations caused by CRISPR or DNA-damaging agents[61-67]. Second, as translational

regulation does not directly interfere with DNA sequences, its risks of carcinogenesis

and off-target effects are markedly reduced, enhancing clinical safety[63]. Critically,

peptide variants generated through translational misreading exhibit high spatiotemporal

heterogeneity[62,66]. These transiently existing non-self-epitopes can effectively evade

immune editing pressures, breaking immune tolerance induced by fixed mutation-

derived epitopes[67]. Consequently, epigenetic regulation and post-translational

modifications have become emerging directions for neoantigen development[63,65].

Studies show that dysregulation of translational fidelity can generate non-canonical

peptides (including antigens from unannotated open reading frames [nuORFs]) via

"misreading" events[62]. Notably, while mechanisms of ribosomal RNA methylation (e.g.,

m4C) in modulating the decoding center are well-characterized in prokaryotes, the

association between such modifications and neoantigen diversity in eukaryotic systems,

particularly the tumor immune microenvironment, remains unknown[68,69].

Our previous studies revealed that METTL5, an m6A methyltransferase catalyzing

adenosine 1832 modification in 18S rRNA, plays a crucial role in tumorigenesis[70].

METTL5 copy number amplification is common across multiple cancers, and METTL5

depletion generates neoantigens through non-canonical frame translation. This

unexpected mechanism expands both the diversity and sources of tumor neoantigens,

offering a novel pathway to enhance anti-tumor immune responses. Based on these

findings, this study integrates multi-omics analysis with functional validation to

demonstrate, for the first time, that METTL5-mediated m6A modification in ribosomes

restricts neoantigen generation by precisely regulating translational fidelity.

Furthermore, we developed a reversible immunotherapeutic strategy targeting this

pathway, providing an innovative solution to overcome tumor immune editing.

RESULTS

METTL5 is identified as a negative regulator of tumor immunogenicity

Based on multi-omics data analysis from the ROCPOT immunogenomics platform, we systematically compared the differential expression profiles of 11 key translation regulatory factors between immune checkpoint blockade (ICB) therapy responders (n=82) and non-responders (n=72). Screening results revealed that multiple translation regulatory factors participate in tumor immune regulation (Fig. 1A). Notably, Fibrillarin (FBL) plays a significant role in tumorigenesis, progression, drug resistance, and immune evasion, with its overexpression typically associated with aggressive tumor phenotypes and poor prognosis[71,72]. Additionally, the expression levels of METTL5 and TRMT112 showed significant negative correlation with ICB treatment response. Our previous research demonstrated that METTL5 must form a functional heterodimeric complex with TRMT112 to catalyze m^{6}A methylation of adenosine at position 1832 in 18S rRNA[73]. This suggests that m^{6}A1832 modification in 18S rRNA may serve as the key molecular mechanism by which the METTL5-TRMT112 complex regulates tumor immunity. These findings indicate that targeting translation regulatory mechanisms may represent a novel strategy for optimizing immunotherapy, offering potential therapeutic avenues to overcome ICB resistance.

METTL5 loss alters tumor growth in syngeneic models

To evaluate the functional role of METTL5 in tumor immunogenicity, we generated *Mettl5*-knockout B16-F10 melanoma and 4T1 breast cancer cell lines using CRISPR/Cas9-mediated gene editing (Fig. S1A). Sanger sequencing confirmed biallelic frameshift mutations at the *Mettl5* locus, with indel events clustered around the gRNA target site (Fig. S1A). Quantitative Reverse Transcription PCR (RT-qPCR) demonstrated significant transcript depletion (>70%) in both Mettl5-KO cell lines (Fig. S1B). Furthermore, high-performance liquid chromatography-mass spectrometry (HPLC-MS) quantification revealed complete abrogation of m^{6}A modification activity in Mettl5-KO cells (Fig. S1C). We first performed in vitro characterization of Mettl5-

KO tumor cells. Proliferation kinetics assessed by real-time cell analysis showed no

significant differences between Mettl5-KO and scrambled control (SCR) in either cell

line (Fig. 1C). Similarly, clonogenic survival assays similarly demonstrated

comparable colony-forming abilities (Fig. S1D, S1E). These findings suggest that the

loss of METTL5 does not impair the proliferative capacity of tumor cells *in vitro*.

To determine the immunological consequences of METTL5 depletion, we established

syngeneic tumor models in immunocompromised (nude) and immunocompetent

(C57BL/6N and Balb/c) mice. METTL5 deficiency did not influence the ability of B16-

F10 cells to form tumors in nude mice, which do not have mature T cells (Fig. 1B).

However, when METTL5 deficient cells were inoculated into syngeneic mouse hosts,

their abilities to form tumors were significantly attenuated by comparison with SCR

cells (Fig. 1D, E). Subsequently, we employed flow cytometry to compare the

proliferation of GFP-labeled Mettl5-KO and mCherry-labeled SCR B16-F10 cells

under both *in vitro* and *in vivo* conditions. The results demonstrated that METTL5

deficient cells exhibited preferential growth suppression compared to control cells (Fig.

1F-H), which is consistent with the aforementioned findings. Collectively, METTL5

loss attenuates tumor growth in immunocompetent mice, indicating an immune-

dependent mechanism for its anti-tumor function.

METTL5 knockout enhances tumor immune cell infiltration

To elucidate the mechanism underlying METTL5 deficiency suppresses tumor growth

specifically in immunocompetent hosts, we systematically characterized immune cell

dynamics within the tumor microenvironment (TME). Immunohistochemical (IHC)

analysis of B16-F10 and 4T1 tumors revealed significantly enhanced CD8α^+ T cell

infiltration in Mettl5-KO lesions compared to SCR controls (Fig. 2A, B). To

quantitatively profile immune populations, we isolated leukocytes from B16-F10

tumors using Percoll gradient centrifugation followed by multicolor flow cytometry.

Mettl5-KO tumors exhibited marked increases in CD8α^+ T cells (Fig. 2C, D),

accompanied by elevated proportions of NK cells and TH cells (Fig. 2C, D). These

findings were corroborated by spleen immunophenotyping, which demonstrated

expanded CTL populations in Mettl5-KO tumor-bearing mice (Fig. 2E, F). To

investigate transcriptional alterations in Mettl5-KO tumors, we performed single-cell

RNA sequencing (scRNA-seq) on freshly isolated tumor tissues (Fig. 2G).

Unsupervised clustering identified major immune subsets including CD8α^+ T cells, B

cells, macrophages, and dendritic cells (Fig. 2H). Notably, the TMEs of Mettl5-KO

tumor contained higher proportions of CD8α^+ T cells and B cells compared to controls

(Fig. 2H-J). However, technical limitations during sample transportation restricted

dataset depth, preventing comprehensive subset analysis. Nevertheless, scRNA-seq

data consistently validated the immunostimulatory phenotype observed through IHC

and flow cytometry, reinforcing METTL5's regulatory role in tumor-immune

interactions.

Inhibition of METTL5 improves immunotherapy

Given that METTL5 inhibition has been shown to enhance tumor-infiltrating capacity

of CD8α^+ T cells, which play a critical role in antitumor immunity[74], we further asked

whether METTL5 targeting could potentiate immunotherapy. To determine whether

METTL5 depletion synergizes with immune checkpoint blockade, we evaluated the

combinatorial efficacy of Mettl5-KO and anti-PD-1 therapy in syngeneic tumor models.

In C57BL/6N mice bearing B16-F10 melanomas, Mettl5-KO tumors exhibited

significantly reduced growth compared to SCR controls following anti-PD-1 treatment

(Fig. 3A). The combination of METTL5 inhibition and anti-PD-1 therapy demonstrated

the most robust antitumor effect, with median survival prolonged by 3 days relative to

single-agent groups (Fig. 3A). These findings were recapitulated in the PD-1-resistant

4T1 breast cancer model. Mettl5-KO 4T1 tumors treated with anti-PD-1 showed

enhanced therapeutic responses, including reduced tumor volumes, and extended

survival compared to other treatment arms (Fig. 3B).

To further evaluate the therapeutic efficacy of combinatorial METTL5 inhibition and

anti-PD-1 blockade, we analyzed tumor-infiltrating immune cells in B16-F10

melanoma tissues using flow cytometry (Fig. 3C, Fig. S3A). The results showed that

METTL5 depletion synergizes with anti-PD-1 treatment to enhance CD8α⁺ T and NK

cell mediated antitumor immunity. Compared to SCR tumors, Mettl5-KO tumors

exhibited higher proportions of CD8α⁺ T, CD4⁺ T cells, and NK cells in both

normalization methods (Fig. 3D, Fig. S3B). Consistent with previous findings[75], anti-

PD-1 therapy expanded tumor-infiltrating cytotoxic immune cell populations, including

CD8α⁺ T cells, and augmented their functional activity (as measured by elevated

CD107a expression). Furthermore, Mettl5-KO synergized with anti-PD-1 treatment to

further enhance these effects (Fig. 3D, S3B). Together, our results reveal that METTL5

knockout potently augments the anti-tumor efficacy of anti-PD-1 treatment, suggesting

that METTL5 inhibition could overcome immune checkpoint inhibitor resistance and

improve therapeutic outcomes in cancer immunotherapy.

224   **METTL5 knockout specifically enhances aberrant protein expression by**
225   **decreasing ribosomal translation fidelity**

METTL5 has been identified as the primary methyltransferase responsible for

catalyzing m⁶A1832 modification on 18S rRNA[70,76,77], its precise molecular

mechanisms and functional implications under physiological and pathological

conditions remain poorly understood, particularly regarding its role in translational

regulation. The A1832 site, catalyzed by METTL5, resides in the ribosomal decoding

center and exhibits remarkable evolutionary conservation. In E. coli, while A1832

remains unmodified, its paired C1402 carries an m⁴Cm modification that is essential

for stabilizing the small ribosomal subunit rRNA structure, facilitating ribosome

assembly, and enhancing translation efficiency and fidelity[68,69]. Higher eukaryotes lack

m⁴Cm1402, suggesting functional compensation by the m⁶A1832 modification at the

corresponding position. Thus, we speculate that METTL5 likely regulates translation

fidelity. To test this hypothesis, we employed a dual-luciferase reporter system. This

system employed mutant plasmids containing either UGA stop codons or ±1 frameshift

mutations downstream of the firefly luciferase start codon, to assess firefly luciferase

mRNA translation efficiency in SCR and Mettl5-KO cells. In the control group (no

mutations), no significant difference was observed in the Firefly/Renilla ratio between

SCR and KO cells, indicating that METTL5 knockout does not inherently disrupt basal

translation efficiency. However, the expression of firefly luciferase in Mettl5-KO cells

was significantly higher than that in SCR cells under mutant conditions (Fig. 4A, B).

Specifically, the UGA stop codon group demonstrated increased ribosomal readthrough

of termination signals, while the -1/+1 frameshift mutation groups revealed a reduced

ability of ribosomes to correct frameshift errors. These results directly support the

notion that METTL5 knockout decreases ribosomal translation fidelity, manifesting as

enhanced tolerance to termination signal readthrough and frameshift errors, thereby

promoting the expression of aberrant proteins encoded by mutant plasmids.

Loss of METTL5 enhances immunogenic neoantigen production via non-

canonical ORF translation

Given the observed translational perturbations in Mettl5-KO cells, we hypothesized

that METTL5 deficiency might impair translational fidelity, potentially leading to

increased translation of non-canonical open reading frames (nuORFs). This could

promote tumor neoantigen generation and consequently enhance tumor

immunogenicity. Unlike conventional gene mutations (which are restricted by central

tolerance mechanisms and thus fail to elicit strong immune responses), nuORFs may

represent more promising novel antigen targets[63,65]. To test this hypothesis, we first

performed hierarchical prediction of translated nuORFs in Mettl5-KO versus SCR

tumor cells using ribosome profiling (Ribo-seq) (Fig. S3A, S4B). The prediction

strategy was as follows (Fig. 4C): The actively translated ORFs were categorized into

annotated ORFs and nuORFs, and nuORFs were further classified into shared nuORFs

(expressed in both groups with a TPM difference of less than 1.5-fold between the two

groups), KO-specific nuORFs (present only in the KO group or with a TPM exceeding

that in the SCR group by 1.5-fold), and SCR-specific nuORFs (present only in the SCR

group or with a TPM exceeding that in the KO group by 1.5-fold). Subsequently, all

nuORFs were categorized into five subtypes[65]: novel ORFs, uORFs (upstream ORFs),

dORFs (downstream ORFs), overlap_uORFs (overlapping upstream ORFs), and

overlap_dORFs (overlapping downstream ORFs) (Fig. 4D). Further analysis revealed

that nuORFs primarily originate from protein-coding genes, pseudogenes, and long

non-coding RNAs (lncRNAs) (Fig. S4C). In contrast to annotated ORFs, nuORF

translation initiation is heavily dependent on non-canonical start codons-particularly

CUG (Fig. S4D)-implying active translation in non-coding genomic regions.

Furthermore, the polypeptides translated from nuORFs are significantly shorter than

those produced by annotated ORFs (Fig. S4E). Analysis revealed that while METTL5

knockout had minimal impact on global translation (Fig. S4F), it significantly enhanced

the production of KO-specific novel ORFs (Fig. 4E, Fig. S4G). Besides, the translation

efficiency of novel nuORFs in the KO group was higher than that in the SCR group

(Fig. S4H), indicating that the extent of translation activation in non-coding regions

was much higher in the KO group than in the SCR group. In Ribo-seq data analysis, if

the 5' end position of ribosome profiling fragment (RPF) reads meets the condition of

being a certain number of base pairs away from the ORF start site such that the base

count modulo 3 equals 1, it indicates that the reading frame is correct (frame 0),

meaning that the ribosome is translating along the mRNA without a frameshift. Further

analysis of reading frame integrity revealed (Fig. 4F) that in the annotated ORF group,

the SCR group had a higher in-frame rate than the KO group, suggesting an increased

probability of translation frameshifts in the KO group. However, in the nuORF group,

the KO group exhibited a significantly higher in-frame rate than the SCR group,

indicating that these KO-specific nuORFs were more consistent with the standard

triplet codon reading frame, which may imply more efficient translation or more precise

start site recognition.

To verify whether the high in-frame KO-specific nuORFs are truly translated and

whether these nuORFs have the potential to become neoantigens presented by MHC-1,

we confirmed the existence of their translation products through MHC-1-IP-MS. Based

on the property that neoantigens require a polypeptide length of at least 8 amino acids,

we selected nuORFs with amino acid lengths greater than 8 as the reference for the IP-

MS database search. Although the number of KO-specific nuORFs was less than that

in the SCR group, HPLC-MS search results showed that 165 translation products of

KO-specific nuORFs were detected, significantly higher than the 68 detected in the

SCR group (Fig. 4G). For example, the peptide sequence NH_2-R-G-P-F-A-D-E-N-F-

COOH identified in Mettl5-KO cells, displayed the hallmark 3-nucleotide periodicity

of ribosomal translation. Subsequent HPLC-MS analysis confirmed the presence of its

amino acid hydrolysis products (Fig. 4H). Together, these findings indicate that

METTL5 loss enhances translation of nuORFs.

METTL5 deficiency enhances TCR clonal diversity in tumors

Based on our previous findings that METTL5-deficient tumors generate increased

neoantigen loads, we hypothesized that METTL5 knockout would drive enhanced T-

cell receptor (TCR) repertoire diversity in tumor-bearing hosts. To test this, we

performed TCR sequencing on peripheral blood samples from B16-F10 melanoma

mouse models at experimental endpoints and conducted comparative TCR repertoire

analysis between Mettl5-KO and SCR groups. CDR3 length distribution histograms

revealed conserved overall patterns between groups, with dominant peaks observed at

10–15 amino acid (aa) lengths (Fig. S4I). Quantitative analysis showed significant

differences in TCR repertoire diversity between Mettl5-KO and SCR tumor-bearing

mice. The Shannon entropy of the KO group (mean = 11.5) was significantly higher

than that of the SCR group (mean = 10.5, $p < 0.05$), indicating greater TCR repertoire

diversity in tumor-bearing hosts with METTL5-deficient tumors (Fig. 4I). Furthermore,

the clonality index was significantly lower in the KO group compared to the SCR group,

suggesting reduced dominance of private clones and a more evenly distributed clonal

landscape in METTL5-deficient tumor-bearing mice (Fig. 4J). Diversity estimations

across Hill diversity profiles (Q values 0–6) consistently demonstrated higher diversity

indices in the KO group (Fig. 4K). Collectively, the above results provide evidences

that METTL5-deficient tumors enhance neoantigen-driven T-cell activation. This

expanded TCR landscape may improve recognition capacity for heterogeneous

neoantigens generated by aberrant translation, potentially strengthening anti-tumor

immunity. Together with our prior demonstration of METTL5-dependent translational

fidelity regulation, these data establish a mechanistic link between ribosomal RNA

modification defects and tumor immunogenicity.

METTL5 depletion unveils targetable neoantigens for vaccine-based tumor

immunotherapy

Building on our prior evidence that METTL5 deficiency enhances MHC-I-restricted

neoantigen presentation, we hypothesized that genetic ablation of β2-microglobulin

(B2M), a critical component of the MHC class I heterodimer, would abolish the

antitumor effects observed in METTL5 deficiency tumors. In order to test this

hypothesis, *B2m* was disrupted using the same CRISPR-Cas9 system as for *Mettl5*

knockout, and the knockout efficiency was validated by Western blotting (Fig. S5A,

S5B). In syngeneic tumor models, we compared the growth kinetics of tumors derived

from *METTL5-B2M* double-knockout cells and *B2M*-knockout scramble control cells

(Fig. 5A). Both groups exhibited comparable tumor growth rates throughout the

experimental period, with no statistically significant differences in tumor volume. In

contrast, the Mettl5-KO single-target group recapitulated the tumor growth suppression

previously observed, confirming the specificity of the antitumor phenotype. Survival

analysis further corroborated these findings. These data definitively establish that

MHC-I-dependent antigen presentation is indispensable for the antitumor immunity elicited by METTL5 depletion.

Based on neoantigen sequences identified through Ribo-seq and MHC-I-IP-MS analyses of Mettl5-KO tumors, we developed peptide-based and mRNA-based vaccines targeting these epitopes. Their therapeutic potential was evaluated in syngeneic B16-F10 tumor models. Longitudinal tumor monitoring revealed that the SCR+Poly(I:C)+peptide group exhibited significantly suppressed tumor growth compared to the SCR+Poly(I:C) control group (Fig. 5B). Survival analysis further demonstrated that median survival extended from 28 days (SCR+Poly(I:C)) to 35 days in vaccinated mice (SCR+Poly(I:C) + peptide), confirming the protective immunity elicited by METTL5-related neoantigens. Similarly, mRNA vaccines encoding these neoantigen sequences showed robust antitumor effects, with the SCR+liposome+mRNA group displaying significantly reduced tumor volumes compared to the liposome-only control (SCR+liposome) (Fig. 5C). However, these therapeutic benefits were not observed in Mettl5-KO tumors, where neither peptide nor mRNA vaccines further improved survival or tumor suppression.Flow cytometry analysis revealed differences in immune cell infiltration(Fig. S5C). Vaccination significantly increased intratumoral CD8$^+$ T cell proportions in control tumors (SCR+liposome+mRNA vs. SCR+liposome), whereas Mettl5-KO tumors showed no CD8$^+$ T cell enrichment regardless of vaccination status (KO+liposome+mRNA vs. KO+liposome). NK cell dynamics followed a similar trend, with vaccinated control tumors exhibiting approximately a two-fold increase in NK cell infiltration compared to controls. In contrast, Mettl5-KO tumors-maintained baseline NK cell levels irrespective of treatment. These observations are consistent with an association between METTL5 deficiency and altered immune cell infiltratin dynamics, though causality remains undetermined.

MATERIALS AND METHODS

Cell culture

The B16-F10 melanoma cells and 4T1 breast cancer cells were purchased from Wuhan Pricella Biotechnology. All cells were authenticated by STR profiling and testing for mycoplasma contamination was performed with a MycAway one-step mycoplasma detection kit (Yeasen, 40612ES25). All cells were cultured in Dulbecco's Modified Eagle Medium (DMEM, Gibco) supplemented with 10% fetal bovine serum (FBS, Biological Industries), 1% penicillin-streptomycin (Thermo Fisher Scientific), and 1% SaveIt mycoplasma inhibitor (Hanbio Biotechnology). Cells were maintained in a humidified incubator at 37°C with 5% CO_2 and routinely tested for bacterial and mycoplasma contamination.

Plasmid construction and Generation of knockout cell

Knockout cells were generated using CRISPR/Cas9 technology. Two independent sgRNAs (targeting *Mettl5* or *B2m* exon 2) were cloned into pL-CRISPR.EFS.GFP plasmid separately. To generate *Mettl5* knockout cells, 2 μg of sgRNA vector that targeting mouse *Mettl5* was transfected to the corresponding cells using Lipofectamine 3000 (Invitrogen). After 2 days' culture in DMEM medium, cells were screened according to fluorescence and isolated into monoclons by flow cytometry. After 3 weeks' culture, colonies were screened out and confirmed by sequencing and RT-qPCR. LCMS/MS was also used to further detected the changes of m^6A modification in SCR and *Mettl5-KO* cells. To generate *Mettl5*-knockout plus *B2m* knockout B16F10 cells, 2 μg of sgRNA vector that targeting mouse *B2m* was transfected to B16 SCR or *Mettl5-KO* cells using Lipofectamine 3000 (Invitrogen). After 2 days' culture in DMEM medium, cells were screened according to fluorescence and isolated into monoclons by flow cytometry. After 3 weeks' culture, colonies were screened out and confirmed by western blot.

sgRNA sequences targeting mouse *Mettl5* locus:

AGCATCGGAGCGGCAATGCT

sgRNA sequences targeting mouse *B2m* locus:

AGTATACTCACGCCACCCACCGG

Cell proliferation and colony formation

For cell proliferation, 1.5×10^5 cells were seeded in a 24-well plate on day 0 and

counted on day one until day 3 with a cell counter (Thermo Fisher Scientific) to measure

cell proliferation. For colony formation, 500 cells were seeded in a 6-well plate and

cultured for six days. Cells were then fixed with 4% formaldehyde (Beyotime) and

stained with 0.1% crystal violet solution (Solarbio). Next, the number and size of Cell

colonies were measured.

Western blot

For western blotting, a six-well plate of cells at approximately 80-90% confluency was

lysed with 100 μl RIPA buffer on ice. Cell lysates were separated by 4-20% SDS-PAGE,

and the protein was then transferred to 0.4 μm polyvinylidene fluoride (PVDF, Merck

Millipore) membrane. After blocking with 5% milk in TBST, the membranes were then

incubated with the corresponding antibodies overnight at 4°C. Membranes were

incubated with secondary HRP-conjugated antibody (Proteintech) at room temperature

for 2 hours following three times washes with TBST and were washed three times again

before detected by the phosphoimager. The corresponding antibodies included

Monoclonal rabbit anti-B2m antibody with 1:5000 dilution (Abcam, ab75853) and

anti-GAPDH antibody with 1:5000 dilution (Proteintech, 10494-1-AP).

Quantitative Reverse Transcription PCR (RT-qPCR)

Total RNA was isolated from B16F10 and 4T1 cells using Total RNA Isolation Kit

(Vazyme) and reverse-transcribed to cDNA by the use of RT reagent Kit (TaKaRa).

Reactions were run on the real-time PCR machine (QuantStudio 7 Flex) using TB

Green PCR master- mix (TaKaRa). The relative gene expression level was calculated

with the comparative cycle threshold (CT) method, and the relative number of target

transcripts was normalized according to mouse GAPDH gene. Each experiment was

performed in triplicate.

METTL5:

Forward primer: 5'- CTGAAGACTGCTTTGGGAATGGC-3';

Reserve primer: 5'- AATGCTGGTAGATCATATCGAAGC-3'.

GAPDH:

Forward primer: 5'-CATCACTGCCACCCAGAAGACTG-3';

Reserve primer: 5'- ATGCCAGTGAGCTTCCCGTTCAG-3'.

Animal Sources

Six- to 8-week-old female C57BL/6N mice were purchased from Beijing Vital River

Laboratory Animal Technology Co., Ltd. Six- to 8-week-old female Balb/c mice and

3- to 4-week-old female nude mice were obtained from GemPharmatech. All mice were

bred in a specific-pathogen-free (SPF) animal facility with a 12-h light/12-h dark cycle,

at $22.5 \pm 1°C$ and $50 \pm 5\%$ relative humidity. All animal experiments were approved

by the Institutional Animal Care and Use Committee (IACUC) of Southern University

of Science and Technology.

In vivo competition assay

For generation of mCherry-labeled SCR and GFP-labeled Mettl5 knockout KO cells,

lentiviral particles were produced by co-transfecting 293T cells with either the GFP or

mCherry plasmid along with the packaging plasmids pMD2.G and PsPAX2. SCR B16

cells were transduced with the GFP-expressing lentivirus, while Mettl5-KO cells were

infected with the mCherry-expressing lentivirus. Following three days of culture, cells

were fluorescence-sorted by flow cytometry to isolate stable populations expressing the

respective markers. A 1:1 mixture of mCherry-labeled SCR cells and GFP-labeled KO

cells was prepared. This mixture was either subcutaneously injected into C57BL/6N

mice or co-cultured in vitro. Tumors were harvested 12 days post-injection, and the

relative proportions of GFP+ (Mettl5-KO) and mCherry+ (SCR) cells in both tumors

and in vitro cultures were quantified by BD Canto II flow cytometer (Flow Cytometry

Shared Facility, SUSTech University).

Tumour growth in mice

Syngeneic and xenogeneic subcutaneous tumor transplantation assays to evaluate

tumorigenicity of mettl5-deficient tumor cells. For syngeneic models, tumor cells were

resuspended in PBS (Gibco) and injected subcutaneously into the right dorsal flank of

C57BL/6N or Balb/c mice. For xenogeneic models, tumor cells were mixed with an

equal volume of Matrigel (Corning) at a 1:1 ratio before subcutaneous injection into

nude mice. Tumor volume was measured every other day starting from day 5 or 7 until

tumors reached a maximum size of 2000 mm^3, calculated as length × width^2. At the

experimental endpoint, tumors were harvested and weighed.

Preparation of tumor tissue single-cell suspension

At the endpoint of the mouse tumor model, tumors were bluntly dissected intact using

surgical scissors and forceps (RWD). After all tumors were stripped, the mass of each

tumor sample was recorded and transferred to a 6-cm dish with appropriate HBSS

(Gibco). Using curved scissors (RWD), tumors were uniformly minced until passable

through a 1-ml pipette tip. Then, collagenase IV (0.5 mg/ml, Sigma) and DNase I (100

U/ml) were added proportionally to the tumor volume. Digestion was performed on a

37°C shaker at 80 rpm for 45 min. The tumor suspension was filtered through a 70-μm

strainer, and single-cell suspension was obtained by centrifugation at 1500 rpm for 10

min at 4°C, followed by three PBS washes. Finally, cells were collected, stained with

antibodies, and analyzed by flow cytometry.

Analysis of tumour-infiltrating lymphocytes

Single-cell suspensions of tumor tissues were prepared and incubated with 50 μL of a solution containing anti-mouse CD16/32 (BioLegend, catalog #101320, diluted 1:50) and anti-mouse Zombie NIR™ Fixable Viability Kit (APC-750, BioLegend, catalog #423105, diluted 1:200) to block non-specific binding and label dead cells. After incubation at 4°C for 10 minutes, a cocktail of surface antibodies for mouse immune cell subset staining was prepared. Cells were stained with the antibody cocktail in 100 μL of tumor staining buffer (BioLegend, #420201) on ice at 4°C for 30 minutes. After incubation, cells were washed twice with staining buffer and resuspended in 900 μL of staining buffer. Counting beads (Invitroge) were added (100 μL per sample) to quantify cell concentrations. Samples were analyzed using a BD Canto II flow cytometer (Flow Cytometry Shared Facility, SUSTech University), with 300,000 CD45$^+$ events recorded per sample. Data were analyzed using FlowJo (version 10).

The antibodies used were as follows: anti-mouse CD4 (Pacific Blue, BioLegend, #100531, diluted 1:200); anti-mouse CD45 (Brilliant Violet 510, BioLegend, #103138, diluted 1:100); anti-mouse CD19 (FITC, BioLegend, #115505, diluted 1:200); anti-mouse CD8a (PerCP/Cyanine5.5, BioLegend, #100734, diluted 1:200); anti-mouse NK-1.1 (PE, BioLegend, #108708, diluted 1:200);anti-mouse CD25 (PE/Dazzle 594, BioLegend, #102048, diluted 1:200); anti-mouse TCR β chain (PE/Cyanine7, BioLegend, #109222, diluted 1:200); anti-mouse CD107a (APC，BioLegend, #121615，diluted 1:200); anti-mouse CD11c (APC/Fire™ 750, BioLegend, #117352, diluted 1:200).

Immunohistochemistry (IHC) detection

IHC was used to evaluate the expression of CD8α in tumor tissues and compare SCR and Mettl5-KO samples. Paraffin-embedded tumor tissues were sectioned into 4-5 μm thick slices and mounted on glass slides. The sections were deparaffinized using xylene and alcohol, followed by antigen retrieval in a citrate buffer. Nonspecific binding was

blocked with an appropriate solution, and then a primary antibody specific to the CD8α

protein was applied and incubated overnight at 4°C. After washing, a secondary HRP-

conjugated antibody was added and incubated for one hour at room temperature. The

bound antibodies were visualized using an IHC detection substrate that produced a

colorimetric reaction. Counterstaining with hematoxylin was performed to highlight

cell nuclei. Images were captured using a microscope, and positive staining cells were

quantified within a defined area using image analysis software. Statistical analysis was

conducted to assess the differences in the number of stained cells between SCR and KO

samples, and the findings were reported with representative images, quantitative data,

and a discussion of the results.

Luciferase Reporter Assay

we employed a dual luciferase reporter assay using the pmirGLO vector, which

contains both firefly (F-Luc) and Renilla (R-Luc) luciferases. The activity of F-Luc,

normalized to R-Luc activity, was used to evaluate the reporter translation efficiency.

Specifically, mutation plasmids were engineered by introducing a UGA stop codon and

frameshift mutations (+1 or -1) into the firefly luciferase gene. In contrast, the wild-

type (WT) plasmid did not contain these mutations, thus maintaining the normal firefly

luciferase sequence. B16-F10 cells, both WT and METTL5 KO, were transfected with

1 μg of either the empty vector or mutation reporter plasmid. After 48 hours post-

transfection, luciferase expression was measured using the Dual-Glo Luciferase Assay

System (Promega).

Ribo-seq

Two 15-cm plates of cells with approximately 80-90% confluency was used per group.

Cells were incubated with 100 μg/ml Cycloheximide (MCE) for 8 min at 37°C and

were then lysed on ice for 10min in 400uL lysis buffer (20 mM Tris–HCl pH 7.5, 150

mM NaCl, 5 mM $MgCl_2$, 1 mM DTT, 100 μg/ml Cycloheximide, 1% Triton X-100).

The supernatant was collected via centrifugation at 12 000 g for 10 min at 4 °C, and its'

absorbance at 260nm was measured. The ribosomes with equal A260 in each group

were reserved for RNA sequencing. Added 15U RNase1 (Thermo Fisher Scientific) for

each 1OD of lysate, and incubated the reactions at 22°C for 40 min with gentle mixing.

Ribosome-protected fragments (RPFs) were collected by MicroSpin S-400 columns

(Cytiva) and purified by Zymo Research RNA Clean & Concentrator-25 kit (Zymo).

The RPFs were then separated with 15% polyacrylamide TBE–urea gel, excised the gel

slices between 17 and 34 nucleotides and recovered RNA from PAGE gel by ZR small-

RNA™ PAGE Recovery Kit (Zymo). Purified RNA was treated with PNK(NEB), and

the RNA library was prepared by VAHTSTM Small RNA Library Prep Kit for

Illumina®(Vazyme). The library concentration was analyzed using the VAHTSTM

Library Quantification Kit for Illumina®, and equimolar pools of libraries were

sequenced.

MHC peptide enrichment

For MHC peptide isolation, four 15-cm plates of B16 SCR or Mettl5-KO cells were

harvested and subsequently lysed on ice in 8 mL of 1× Lysis buffer (comprising 0.5%

NP-40, 50 mM Tris pH 8.0, 150 mM NaCl, and protease inhibitor) for 10 minutes.

Following lysis, protein extracts were obtained by centrifugation at 12,000 g for 10

minutes at 4°C. The supernatants were then incubated overnight with anti-mouse MHC-

I antibody conjugated to Pierce Protein A/G magnetic beads (Santa Cruz) to facilitate

immunoprecipitation. The bead-protein mixture was first treated with 1 mL of Wash

Buffer 1 (0.5% NP-40, 50 mM Tris-HCl pH 8.0, 150 mM NaCl, 5 mM EDTA, 100 μM

PMSF, and protease inhibitor) to remove nonspecifically bound proteins while

simultaneously inhibiting proteolytic degradation. Subsequent washes included 1 mL

of Wash Buffer 2 (50 mM Tris-HCl pH 8.0, 150 mM NaCl) to eliminate residual

detergent from the first step. High-affinity contaminants were then dissociated using 1

mL of Wash Buffer 3 (50 mM Tris-HCl pH 8.0, 450 mM NaCl), which applied high

ionic strength to disrupt non-specific interactions. Finally, a salt-free wash with 1 mL

of Wash Buffer 4 (50 mM Tris-HCl pH 8.0) ensured complete removal of residual salts,

thereby maintaining compatibility with downstream elution steps. Bound peptides were

released by treating the beads with 10% acetic acid (pH ~3.0), which denatures MHC

molecules and facilitates peptide dissociation. The eluate was then filtered through a 3

kDa molecular weight cut-off filter to remove the MHC-I α-chain, B2m protein, and

the corresponding MHC antibody. The eluted proteins were ultimately analyzed by

HPLC-MS to identify the MHC peptides. Acid elution was performed on six 15 cm

dishes of B16 control cells or B16- Mettl5-KO cells stimulated with 100IU/ml IFN-γ

for 24h. Cells were treated on ice with pH3.3 citric acid-phosphate buffer for 5min to

dissociate MHC-antigen peptide complexes and release antigenic peptides. After

centrifugation at 10000g, the supernatant was collected and filtered through a 3 kDa

molecular weight cutoff filter. The eluted proteins were analyzed by HPLC-MS for

MHC peptides.

TCRβ library construction and TCRβ sequencing

PBMCs were isolated from whole blood of C57BL/6N mice that had been inoculated

with either control B16F10 cells or Mettl5-KO cells 12 days prior. DNA was extracted

from PBMC using a QIAamp DNA Blood Mini Kit (Qiagen), and the concentration

was tested using NanoDrop 2000 spectrophotometer (Thermo Fisher Scientific). The

DNA was used as a template for PCR amplification, which was performed to generate

the library of TCR and BCR. The step 1 PCR amplification protocol was as follows:

95 °C for 5 min; 95 °C for 30 s, 59 °C for 30 s; 72 °C for 1 min for 30 cycles; and 72 °C

for 10 min. The step 2 PCR amplification protocol was as follows: 98 °C for 2 min;

98 °C for 30 s, 65 °C for 30 s; 72 °C for 30 s for 10 cycles; and 72 °C for 5 min. The

PCR products were purified, and the barcodes were confirmed. The annealed region of

these primers was the CDR3 region, as shown by Multiplex PCR Amplifies. The CDR3

regions were sequenced on Illumina NextSeq500 (MyGenostics). According to the

sequencing depth, the test covered about 90% of the TCR. The CDR3 regions were

identified based on the definition established by the International ImMunoGeneTics

(IMGT) (http://www.imgt.org/, accessed on 19 March 2021) collaboration, and the V,

D, and J segments contributing to each CDR3 region were identified by a standard

algorithm. Data were analysed using ImmunoSEQ Analyzer 3.0, an Adaptive

Biotechnologies online analysis platform.

RNA-seq and ribo-seq analysis

Adapter sequences were removed from both Ribo-seq and RNA-seq reads using

TrimGalore (v0.6.7). For Ribo-seq data, reads shorter than 20 nucleotides or longer

than 35 nucleotides were discarded to ensure high-quality ribosome-protected fragment

selection. The trimmed reads were then aligned to rRNA and tRNA sequences using

Bowtie2 (v2.2.5)[78]. Reads that aligned to rRNA or tRNA were excluded, while the

remaining unaligned reads were retained for subsequent analysis. Processed reads were

aligned to the reference genome (gencode GRCh38) by the STAR (v2.7.10a)[79],

allowing for at most two mismatches. For Ribo-seq data, quality control and

translational profiles was analyzed by RiboseQC. Transcriptome assembly and

quantification were performed using StringTie (v2.1.7)[80]. Both read counts of Ribo-seq

and RNA-seq were converted to TPM (transcripts per million). Genes with sufficient

expression level (Ribo TPM > 1 and RNA TPM > 5) were subjected to further analysis.

Translation efficiency (TE) and the differentially-TE genes (DTEGS) were calculated

by RiboDiff (v0.2.2)[81], GO term analysis were performed by Metascape

(https://metascape.org/gp/index.html). GraphPad Prismsoftware and R was used for

data presentation.

Hierarchical prediction of translated ORFs across tissues

In our study, we used the RiboCode[82] algorithm to detect actively translated open

reading frames (ORFs) with the parameters set as follows: a minimum fractional

threshold of ribosome profiling reads (-f0_percent 0.5), a primary start codon of ATG

(--start_codon ATG), alternative start codons of CTG, GTG, and TTG (--

alt_start_codons CTG, GTG, TTG), and a minimum amino acid length of 8 (--min-AA-

length 8). The identified ORFs were classified into several categories: "annotated"

ORFs, which overlap with annotated coding sequences (CDS) and share the same stop

codon; "uORFs", located upstream of the annotated CDS without overlap; "dORFs",

situated downstream of the annotated CDS without overlap; "Overlap_uORFs",

overlapping with the annotated CDS while being upstream; "Overlap_dORFs",

overlapping with the annotated CDS while being downstream; "Internal" ORFs, which

are internal to the annotated CDS but in a different reading frame; and "novel" ORFs,

originating from non-coding genes or non-coding transcripts of coding genes.

Additionally, we generated collapsed BED12 files for annotated ORFs

(annotatedORFdb.bed) and non-canonical ORFs (excluding the "annotated" type,

nuORFdb.bed), which served as references for subsequent analysis.

Generating nuORFdb v.1.0

To count the number of ribosomal profiling (RPF) reads aligned to ORFs, we first

corrected the read positions based on the offset distances between the 5' end of the reads

and the ribosomal A-sites. Subsequently, we employed the "rpfTPM" script from a

prior study (available at link) to quantify the number of in-frame reads. This script

normalizes the in-frame read counts to transcripts per million (TPM) against the

nuORFdb.bed file. ORFs with TPM values 1.5-fold higher in the KO samples compared

to SCR samples were designated as "KO-larger nuORFs". Conversely, ORFs with TPM

values 1.5-fold higher in SCR samples relative to KO samples were classified as "SCR-

larger nuORFs". Please note that the link to the script could not be parsed due to

network issues. If needed, please verify the link's validity and try accessing it again. If

you don't require the content from this link, the rest of your request has been addressed

as follows.

Estimation of absolute translation levels

UniProt (reviewed mouse, release 2025_02) and the translated amino acid sequences of larger nuORFs were used as reference databases for protein and peptide identification. Notably, only those nuORFs longer than 8 aa were defined as potential translated regions. The proteomics search was performed using SearchGUI (v4.2.17) (https://github.com/compomics/searchgui) with MS-GF+ and Tide as search engines. Carboxyamidomethylation of cysteine was set as a fixed modification, while oxidation of methionine and proline were set as variable modifications. The enzyme specificity was set to 'unspecific' and the precursor mass tolerance was set to 20 ppm Hydrophobicity values for total PSM peptides were predicted with R package protViz (v0.7.9) (https://github.com/cpanse/protViz).

Statistical analysis

No sample size calculations and blinding were performed. There was no method of randomization. No samples or animals were excluded from analysis. Prism 9 software (GraphPad) was used to analyze data from biological experiments by performing two-tailed unpaired Student's t-tests. One-way analysis of variance (ANOVA) with Tukey's test was used to compare multiple groups and two-way ANOVA was used to compare cell proliferation analysis and tumor growth curves. *P* values *$P<0.05$, **$P<0.01$, ***$P<0.001$, ****$P<0.0001$ were considered statistically significant.

DISCUSSION

This study reveals METTL5 as a central ribosomal translation regulator in tumor immune evasion through integrated multi-omics analysis and functional validation. METTL5 depletion markedly enhances tumor neoantigen burden by activating noncanonical upstream open reading frame (nuORF) translation. This mechanism fundamentally differs from traditional mutation-derived neoantigens, as its origin relies not on genomic mutation accumulation but dynamically expands antigenic diversity through translational fidelity regulation, providing a novel strategy to overcome immunotherapy resistance[63,65,67].

At the molecular level, METTL5-mediated m⁶A1832 modification resides at the h44-h45 helical junction of 18S rRNA, a critical site within the ribosomal decoding center responsible for maintaining codon-anticodon pairing accuracy[70,73,76]. Although m⁴C modifications in prokaryotic decoding centers are well-established regulators of translational fidelity[68,69], this study identifies METTL5-catalyzed m⁶A modification at the eukaryotic homologous region as a potential functional analog. Cryo-electron microscopy analysis detected two distinct conformational states in the decoding center of METTL5-deficient 293T cells. However, how these structural alterations quantitatively affect translational fidelity in tumor cells remains unclear. Dual-luciferase reporter assays demonstrated increased read-through rates of UGA stop codons and significant elevation of ±1 frameshifting events in METTL5-knockout tumor cells. Ribosome profiling further confirmed enhanced translation initiation at noncanonical start codons in METTL5-deficient cells, with a notable increase in translated peptide lengths exceeding eight amino acids, thereby generating an expanded neoantigen reservoir for MHC-I presentation.

In terms of tumor immune microenvironment, METTL5-depleted tumors exhibited a pronounced "hot phenotype". Flow cytometry and single-cell sequencing revealed increased CD8α⁺ T cell infiltration density, elevated NK cell proportions, and a T cell receptor (TCR) repertoire with a Shannon diversity index of 11.5 compared to 10.5 in

controls. Notably, TCR clones from METTL5-knockout tumors showed CDR3 regions

preferentially lengthened to 12-14 amino acids, a range associated with enhanced

antigen-binding flexibility[83]. In anti-PD-1 combination therapy models, METTL5

knockout combined with PD-1 blockade significantly prolonged the median survival of

tumor-bearing mice while increasing the proportion of CD107a$^+$ cytotoxic T cells and

NK cells within the tumor microenvironment[75], demonstrating that translational

reprogramming synergizes with checkpoint inhibition to reverse immune exhaustion.

Despite these therapeutic implications, clinical translation of METTL5-targeting

strategies faces multiple challenges. First, METTL5 depletion did not affect in vitro

proliferation of murine melanoma or breast cancer cells, yet its functional context may

exhibit tissue- or species-specificity requiring further investigation. Second, reduced

translational fidelity risks aberrant self-antigen release; although METTL5 is

overexpressed in tumors, long-term autoimmunity monitoring remains necessary. Third,

nuORF-based vaccine efficacy is constrained: in B16-F10 models, MHC-I

immunopeptidome analysis using H2-K^b and H2-D^b antibodies only successfully

enriched peptides via H2-D^b, limiting comprehensive neoantigen identification; acid

elution partially compensates but nuORF translation efficiency remains low with

unverified immunogenicity. Additionally, the sensitivity threshold of mass

spectrometry techniques and database dependency may lead to missed detection of

noncanonical peptides[37].

Future studies should focus on resolving dynamic conformational changes in the

ribosomal decoding center following METTL5 deletion through cryo-electron

microscopy to elucidate atomic-level regulatory mechanisms. Development of specific

METTL5 inhibitors is required to balance translational stress thresholds against

proteostasis disruption. Machine learning platforms should be constructed to prioritize

highly immunogenic nuORF-derived neoantigens. This work not only establishes

ribosomal chemical modifications as critical mediators of tumor immune surveillance

but also provides experimental evidence for "translational reprogramming" as a novel

strategy to overcome current limitations in cancer immunotherapy.

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CONFLICT OF INTEREST

The authors declare no conflicts of interest.

AUTHOR CONTRIBUTIONS

H.C. designed and conceived the experiments. Y.Y.Z., X.Y.S., R.Q.W. and Y.L.C. performed experiments. Y.C.W. and X.Y.S. analyzed the data. Y.Y.Z., X.Y.S. and H.C. wrote the manuscript. All authors have read and approved the final manuscript.

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FIGURE LEGENDS

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Fig. 1: METTL5 depletion attenuates tumour growth

A, Relationship between the expression levels of certain RNA modification enzymes

and response to immune checkpoint blockade (ICB) therapy. METTL5 and TRMT112

are highlighted with purple boxes.

B, Growth curve of tumors in nude mice subcutaneously implanted with B16-F10 SCR

or METTL5 deficiency cells. The tumor growth (left) and final tumor weights (right)

were compared.

C, In vitro proliferation assay of SCR and Mettl5-KO cells in B16-F10 (left) and 4T1

(right) cell lines.

D, Growth curve of tumors in C57BL/6N mice subcutaneously implanted with B16-

F10 SCR or Mettl5-KO cells. The tumor growth (left) and final tumor weights (right)

were compared.

E, Growth curve of tumors in Balb/c mice subcutaneously implanted with 4T1 SCR or

Mettl5-KO cells. The tumor growth (left) and final tumor weights (right) were

compared.

F-H, In vivo competition assay between mCherry-labeled SCR control and GFP-

labeled Mettl5-KO B16-F10 cells. Schematic diagram of the competitive growth assay

(**F**). mCherry-labeled SCR and GFP-labeled-KO cells were mixed at a 1:1 ratio and

either subcutaneously injected into C57BL/6N mice or co-cultured in vitro. Cell ratios

were quantified by flow cytometry on Day 12. Representative flow cytometry plots

showing the composition of SCR and Mettl5-KO cells (**G**). Quantitative analysis of cell

population ratios (**H**).

Note: A, Data from The Cancer Genome Atlas (TCGA) and were presented as median

(IQR 25-75%, n=154). Student's t-tests. **B**, **D**, two-way ANOVA (left) and One-way

ANOVA with Tukey's test (right). **C**, two-way ANOVA. **E**, two-way ANOVA (left)

and Student's t-tests (right). **H**, Student's t-tests. **B**, **D**, **E**, Data were presented as mean

± SEM. **C**, **H**, Data were represented as mean± SD (n=3). *$P<0.05$, **$P<0.01$,

***$P<0.001$, ns: not significant.

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sFig. 1: Genetic ablation of METTL5 via CRISPR-Cas9 and its effect on tumour-cell growth in vitro

A, Schematic diagram of sgRNA targeting mouse *Mettl5* locus. Sanger sequencing was used to confirm CRISPR-Cas9-mediated knockout.

B, HPLC-MS/MS analysis of m^{6}A/A levels in total RNA purified from the SCR and METTL5 KO cells.

C, Knockout of *Mettl5* in B16-F10 cells and 4T1 cells were verified by Real-time quantitative PCR analysis.

D, **E**, Colony-formation assay of WT and METTL5 KO B16-F10 cells. 500 cells were seeded in a 6-well plate and cultured for six days, representative pictures (**D**) were taken, and the colony numbers and sizes (**E**) were measured.

Note: C, One-way ANOVA with Tukey's test. **E**, Student's t-tests (right). **C**, **E**, Data were represented as mean ± SD (n=3). ****$P<0.0001$, ns: not significant.

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Fig. 2: METTL5 deficiency enhances tumor immune cell infiltration

A, B, Immunohistochemistry (IHC) detected CD8α⁺ cells in Mettl5-KO tumor tissues

derived from B16-F10 (**A**) and 4T1 (**B**) models. Representative images (left) and

quantified CD8α$^+$ cell counts (right) are shown. 10 random 20× fields per sample were

analyzed. CD8α$^+$ cells are indicated by red arrows.

C-F, Flow cytometry analysis of immune cell populations in the tumor

microenvironment (**C**) and spleen (**E**) of C57BL/6N mice inoculated with SCR or

Mettl5-KO B16-F10 cells, and representative flow cytometry image showing the

distribution of CD4$^+$ T cells, CD8$^+$ T cells (left) and NK cells(right). The relative

frequency of immune cells amongst living cells from tumors (**D**) and spleen (**F**) were

assessed.

G, Workflow schematic of single-cell sequencing for tumor tissues. In vivo tumor

growth kinetics in C57BL/6 mice inoculated with SCR or Mettl5-KO B16-F10 cells.

Tumors were excised and dissociated at days 18 post-inoculation, followed by CD45$^+$

immune cell enrichment and single-cell sequencing.

H, Dot plot illustrates the clustering analysis results of single-cell sequencing data (left).

T cells were isolated from the total immune cell population and further stratified into

CD8$^+$ T cells and CD4$^+$ T cell subsets (right). Dot color indicates expression level and

dot size indicates the proportion of each cell type expressing each gene.

I, t-SNE plots show T-cell subpopulations derived from tumors, including CD4$^+$ T-cell

and CD8$^+$ T-cell subpopulations.

J, Bar chart shows the proportion of tumor - infiltrating immune cell populations, with

blue representing the METTL5- knockout group and red the control group.

Note: A, **B**, One-way ANOVA with Tukey's test. **D**, **F**, Student's t-tests (right). **A**, **B**,

D, **F**, Data were represented as mean ± SD. *$P<0.05$, **$P<0.01$, ****$P<0.0001$, ns: not

significant.

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Figure. Inhibition of Mettl5 increases the efficacy of tumor immunotherapy

A, Anti-PD-1 antibody was administered to Mettl5-KO and SCR B16-F10 tumor-bearing C57BL/6N mice on days 4, 7, 10, and 13 post-implantation, followed by longitudinal monitoring of tumor volume (left) and overall survival (right).

B, Anti-PD-1 antibody was administered to Mettl5-KO and SCR B16-F10 tumor-bearing Balb/c mice on days 4, 7, and 10 post-implantation, followed by longitudinal monitoring of tumor volume (left) and overall survival (right).

C, t-SNE analysis of immune cell populations in the tumor microenvironment of tumor in C57BL/6N mice. Mice were treated with either isotype control or anti-PD-1 antibody. The analysis includes the following immune cell populations: Lymphocytes, CD45+ cells, Myeloid cells, T Cells, B Cells, NK Cells, CD107a+ cells, CTL, and T Helper (TH) cells. The plots show the distribution of these immune cell populations in the tumor microenvironment for SCR and KO groups treated with isotype control or anti-PD-1 antibody.

D, Quantitative analysis of immune cell populations in the tumor microenvironment of

tumor in C57BL/6N mice. The number of CD4$^+$ T cells, CD8$^+$ T cells, NK cells, and

CD107 a + CD8$^+$ T cells per milligram of tumor tissue was determined by flow

cytometry.

Note: Mouse survival curves were plotted using endpoints: death or tumor volume

exceeding 2000 mm³. **A**, **B**, two-way ANOVA. Data were presented as mean ± SEM.

D, Student's t-tests. Data were represented as mean ± SD (n=5). *P<0.05, **P<0.01,

***P<0.001, ****P<0.0001, ns: not significant.

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sFig. 3: Depletion of PTRHD1 enhances intra-tumoral T-cell infiltration

A, **B**, Representative flow cytometry image (**A**) showing the distribution of CD4$^+$ T

cells, CD8$^+$ T cells (left), CD107a+ CD8$^+$ T cells (middle), and NK cells(right). The

relative frequency of immune cells amongst living cells (**B**) were assessed.

Note: B, Student's t-tests. Data were represented as mean ± SD (n=5). *P<0.05,

**P<0.01, ns: not significant.

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Fig. 4: Depletion of METTL5 induces neoantigen generation

A, Scheme of reporter vector. Dual-luciferase plasmid constructs were designed with a

UGA stop codon, +1 frameshift, or -1 frameshift upstream of the firefly luciferase gene,

with Renilla luciferase as an internal control.

B, Translation efficiency of firefly luciferase in SCR and METTL5 KO cells was

revealed by a reporter assay. Firefly luciferase activity was normalized to Renilla

luciferase, with the SCR group set as the baseline (relative ratio = 1).

C, Workflow schematic of nuORFdb v.1.0 generation. ORFs from ribosome profiling (Ribo-seq) data were classified into annotated ORFs and novel ORFs (nuORFs), which were compiled as a database for subsequent analyses. For differential nuORF expression between Mettl5-KO and SCR groups, we classified nuORFs into three categories: (1) Shared nuORFs (expressed in both groups with <1.5-fold TPM difference), (2) Mettl5-KO -specific nuORFs (exclusive to Mettl5-KO or with ≥1.5-fold higher TPM than SCR), and (3) SCR-specific nuORFs (exclusive to SCR or with ≥1.5-fold higher TPM than Mettl5-KO).

D, Pie chart illustrating the classification of nuORFs identified in Mettl5-KO and SCR group. The chart shows the percentage distribution of different nuORF categories relative to the total nuORFs.

E, Differential analysis of nuORF categories between Mettl5-KO and SCR groups.

F, Distribution of in-frame translation events in Mettl5-KO and SCR groups across annotated ORFs and nuORFs.

G, Peptides eluted from MHC-I immunoprecipitates in SCR and Mettl5-KO samples were identified using corresponding nuORFs (>8 aa) as reference sequences.

H, A representative nuORF-derived peptide NH2-R-G-F-P-A-D-N-F-COOH) in Mettl5-KO tumors. The characteristic 3-nucleotide periodicity of ribosome-protected fragments confirmed its translational activity.

I, Scatter plots depict comparative distributions of TCR repertoire diversity (Shannon entropy) and clonality between Mettl5-KO and SCR groups.

I, Comparative diversity analysis (Hill numbers q=1–6) between Mettl5-KO and SCR groups.

J, Violin plots showing the distribution of clonal expansion levels in Mettl5-KO versus SCR groups, with clones categorized by abundance thresholds: Rare (<1e-05), Small (1e-04), Medium (0.001), Large (0.01), and Hyperexpanded (1).

1151 **Note: B**, **F**, **I, K**, Student's t-tests. **B**, **I,** Data were represented as mean ± SD. **F**, **K**,

1152 Data were presented as median (IQR 25-75%). *$P<0.05$, ****$P<0.0001$, ns: not

1153 significant.

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sFig. 4: Identification of novel ORFs induced by METTL5 deficiency

A, Quality control of Ribo-seq data in SCR (left) and METTL5 KO (right) cells.

B, The PCA plot shows the separation between Mettl5-KO and SCR groups based on

gene counts

C, Distribution of transcript types for nuORFs identified by Ribo-seq.

D, Distribution of start codons for nuORFs and annotated ORFs in Mettl5-KO and SCR

groups.

E, Protein length distribution (in amino acids) for different ORF categories in Mettl5-

KO and SCR groups.

F, Volcano plot depicting down-regulated (blue) and up-regulated (red) genes

identified by Ribo-seq in METTL5 KO cells compared with SCR cells.

G, Differential counts of nuORF types between Mettl5-KO and SCR groups. Red bars

indicate KO-specific nuORFs (present only in the KO group or with a TPM exceeding

that in the SCR group by 1.5-fold), while blue bars indicate SCR-specific nuORFs

(present only in the SCR group or with a TPM exceeding that in the KO group by 1.5-

fold).

H, Translation activity analysis of different ORF categories in Mettl5-KO and SCR

groups. Higher PME values indicate more active translation. The dashed line at PME =

0.5 represents a threshold for translation efficiency.

I, Scatter plot from TCR-seq quality control showing no significant differences in

CDR3 length distribution between KO and SCR groups.

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1178   **Fig. 5: Combination of neoantigen vaccination and METTL5 inhibition elicits**

1179   **robust antitumour immune response**

1180   **A**, Growth curve of tumors in C57BL/6N mice subcutaneously implanted with SCR,

1181   Mettl5-KO, SCR plus B2M-KO, Mettl5-KO plus B2M-KO B16-F10 cells. The tumor

1182   growth (left) and overall survival (right) were compared.

B, C57BL/6 mice bearing B16-F10 tumor were immunized with selected neoantigen-encoded peptides or the adjuvant poly(I:C) on days -3, 3, 7, and 11. The tumor growth (left) and overall survival (right) were compared.

C, C57BL/6 mice subcutaneously implanted with SCR or Mettl5-KO B16-F10 cells were immunized with the neoantigen mRNA vaccination and liposome on days 3, 6, 9, and 12. The tumor growth (left) and overall survival (right) were compared.

D, Working model for the critical role of METTL5 in tumor immunity through neoantigen generation. METTL5 deficiency abrogates m^{6}A deposition in the ribosomal decoding center, impairing ribosomal function and compromising translation fidelity. These defects lead to aberrant non-canonical translation events that generate immunogenic neoantigens, thereby stimulating anti-tumor immunity.

Note: Mouse survival curves were plotted using endpoints: death or tumor volume exceeding 2000 mm³. **A**, **B**, **C**, Student's t-tests. Data were presented as mean ± SEM. *$P<0.05$, ****$P<0.0001$, ns: not significant.

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sFig. 5: Generation of B2M knockout cell lines

A, Schematic diagram of sgRNA targeting mouse *B2M* locus.

B, Western blotting confirmed the depletion of B2M in SCR and Mettl5-KO B16-F10 cells. GAPDH was used as internal control.

C, Flow cytometry analysis of immune cell infiltration in tumors of C57BL/6N mice bearing SCR or Mettl5-KO B16-F10 cells following neoantigen mRNA-liposome vaccination. The frequencies of CD4+ T cells and CD8+ T cells were quantified within live cell populations.

Note: C, Student's t-tests. Data were presented as mean ± SD. *$P<0.05$, **$P<0.01$, ns: not significant.

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