



Prognostic value and immune regulatory role of dynamin 1-like in lung adenocarcinoma

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Background: Lung adenocarcinoma (LUAD) is the most common histological subtype of non-small cell lung cancer (NSCLC), with poor treatment outcomes worldwide. Dynamin-related protein 1 (DRP1), which is encoded by the dynamin 1-like (*DNM1L*) gene, acts as a regulator of mitochondrial fission and plays crucial roles in tumor initiation and progression. However, the clinical value and immune regulation of *DNM1L* in LUAD have not been explored.

Methods: We comprehensively analyzed the expression of *DNM1L* in the LUAD cohort of the Human Protein Atlas (HPA) and the University of The ALabama at Birmingham CANcer data analysis Portal (UALCAN) databases. Kaplan-Meier plotter, in addition to the PrognoScan database, was used to estimate the correlation between *DNM1L* expression and survival outcome of LUAD patients. The association between the immune tumor microenvironment (TME) and *DNM1L* expression in LUAD was evaluated based on the Tumor IMMune Estimation Resource (TIMER)2.0 database. Finally, the functions of *DNM1L* were validated *in vitro* experiments, including reverse transcription-quantitative polymerase chain reaction (RT-qPCR), western blot, wound healing assays, and transwell assays.

Results: *DNM1L* was overexpressed in LUAD compared to healthy control tissues and was regarded as an independent prognostic factor. Overexpression of *DNM1L* was significantly related to clinical variables and poor survival outcomes of LUAD patients. Moreover, *DNM1L* expression was positively associated with the expression of key genes involved in the regulation of immune cell subsets, including T helper (Th)2 cells, Th cells, B cells, CD8 T cells, dendritic cells, and mast cells. In contrast, *DNM1L* was negatively correlated with the infiltrating levels of myeloid dendritic cells and B cells. Furthermore, *DNM1L* may play a role in regulating immune cell infiltration and have prognostic value in LUAD patients. Finally, the *in vitro* experiments showed that increased *DNM1L* significantly promoted the proliferation and migration of

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LUAD cells.

Conclusions: This study suggested that *DNMIL* may play an important role in regulating the proliferation and migration of LUAD cells as well as the infiltration of tumor-related immune cells, which suggests *DNMIL* was a potential therapeutic target in LUAD. Further studies are however warranted to define its exact mechanism of action and potential therapeutic significance in LUAD patients.

Keywords: Dynamin 1-like (*DNMIL*); lung adenocarcinoma (LUAD); immune; prognosis; biomarker

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Introduction

Lung cancer remains the major cause of cancer-related mortality all over the world, with 1.8 million new cases diagnosed and 1.6 million deaths each year. Globally, the incidence of lung cancer continues to rise annually and 5-year survival rate following diagnosis remains at less than 20% (1). Lung adenocarcinoma (LUAD) retains the most common histological subtype of non-small cell lung cancer (NSCLC) and is often diagnosed at an advanced stage, where surgical resection with curative intent is not generally a treatment option for these patients (2). The most recent use of immunotherapies, such as programmed cell death-1 (PD-1), programmed cell death-ligand 1 (PD-L1), cytotoxic

T-lymphocyte antigen 4 (CTLA-4) immune checkpoint inhibitors (ICIs), either in first- or second-line setting, have revolutionized the treatment landscape of NSCLC and has led to significant durable responses in many patients (3,4). However, resistance to ICIs is a common phenomenon, leading to disease progression following an initial response to treatment (5,6). A greater understanding of the biological mechanisms underlying the resistance to ICIs may contribute to the development of more effective immunotherapies (7). One possible mechanism that plays in the development of resistance to ICIs is the inhibition of migration and infiltration of T cells to the immune tumor microenvironment (TME), which results in a functionally exhausted phenotype (8). Thus, identifying the underlying biological mechanisms of T cell exhaustion is crucial in improving the response to immunotherapy. The immune cell infiltration in the TME is particularly relevant to the prognosis and chemosensitivity of patients, in particular neutrophils and tumor-associated macrophages (TAMs) (9). Therefore, identifying the tumor immunophenotype and further understanding of the underlying biological mechanisms of immune regulation are warranted to identify novel immunotherapeutic targets in LUAD (10).

The mitochondria play essential roles in a number of essential cellular functions, such as cell growth, proliferation, differentiation, metabolism, and programmed cell death (11). It is now known that mitochondrial dynamics are implicated in the regulation of fundamental processes related to T cell and natural killer (NK) function and tumor immuno-surveillance (12,13). In fact, mitochondrial morphology may directly regulate T cell differentiation through metabolic pathways. Dynamin 1-like (*DNMIL*)/dynamin-related protein 1 (DRP1) plays important roles in regulating mitochondrial fission and in maintaining the mitochondrial dynamic network, which is important in T cell proliferation, migration, and invasion (14).

Highlight box

Key findings

- Dynamin 1-like (*DNMIL*) was overexpressed in lung adenocarcinoma (LUAD) compared to healthy control tissues and significantly promoted the proliferation and migration of LUAD cells, which suggests *DNMIL* was a potential therapeutic target in LUAD.

What is known and what is new?

- Dynamin-related protein 1, which is encoded by the *DNMIL* gene, acts as a regulator of mitochondrial fission and plays crucial roles in tumor initiation and progression.
- *DNMIL* may play an important role in the recruitment of immune cells in LUAD, potentially contributing to the regulation of immune-related signaling pathways and the infiltration of tumor-related immune cells.

What is the implication, and what should change now?

- *DNMIL* plays a role in regulating the immune infiltration of immune cell subsets in the tumor microenvironment of lung tumors. Further studies are however warranted to define its exact mechanism of action and potential therapeutic significance in LUAD patients.

The molecular mechanisms whereby *DNM1L* regulates tumor progression and the immune TME in LUAD remain largely unknown. Herein, we evaluated *DNM1L* expression in LUAD and the correlation between *DNM1L* and clinical variables based on publicly available databases. In addition, we comprehensively analyzed the correlation between *DNM1L* expression and survival outcomes of LUAD patients based on Kaplan-Meier plotter and PrognScan databases and the association between *DNM1L* expression and the immune infiltration of cells in the TME using the Tumor IMMune Estimation Resource (TIMER)2.0 database. This study provides novel insights into the potential clinical value of *DNM1L* in LUAD and highlights, at least in part, the mechanistic pathways whereby *DNM1L* affects immune cell-tumor interactions. We present this article in accordance with the REMARK reporting checklist (available at <https://tcr.amegroups.com/article/view/10.21037/tlcr-23-685/rc>).

Methods

University of The ALabama at Birmingham CANCER data analysis Portal (UALCAN)

UALCAN (<https://ualcan.path.uab.edu/index.html>) is a comprehensive and interactive online resource for the in-depth analysis of gene expression data based on The Cancer Genome Atlas (TCGA) (15). We evaluated the expression of *DNM1L* in the “lung adenocarcinoma” dataset using the “Expression Analysis” module. Moreover, we also explored the phosphoprotein expression profiles of *DNM1L* in LUAD and adjacent normal lung tissues according to clinical variables, including race, age, tumor grade, and cancer stage, in the Clinical Proteomic Tumor Analysis Consortium (CPTAC) database. The prognostic analysis was performed using the Student’s *t*-test. Statistical significance was determined as a P value <0.05. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

Human Protein Atlas (HPA)

The HPA dataset (<https://www.proteinatlas.org/>) aims to map human proteins based on proteomics, transcriptomic, and systems biology data (16). It includes data on protein expression in normal and tumor tissues, in addition to their prognostic relevance in the form of survival curves. It also contains the survival curve data of tumor patients. *DNM1L*

protein expression in LUAD and healthy control tissues was evaluated by immunohistochemistry (IHC) and scoring of stained tissues based on staining intensity (negative, weak, moderate, or strong), fraction of stained cells, and subcellular localization (nuclear and/or cytoplasmic/membranous).

Gene expression profiling interactive analysis (GEPIA)

The GEPIA database (<http://gepia2.cancer-pku.cn/#index>) offers a tool for the analysis of gene expression data related to cancer pathological subtypes, driver genes, alleles, differentially expressed or carcinogenic factors, so as to explore new cancer therapeutic targets and biomarkers based on normal and tumor samples in the TCGA and Genotype-Tissue Expression (GTEx) databases (17). The “Correlation Analysis” module was used to explore the correlation between *DNM1L* and other genes in the GEPIA database. We analyzed the correlation of *DNM1L* with other genes with spearman coefficient using the non-log scale for calculation and the log-scale axis for visualization. Statistical significance was determined as a P value <0.05.

PrognScan

PrognScan (<http://dna00.bio.kyutech.ac.jp/PrognScan/index.html>) is a tool for evaluating the biological correlation between prognosis and gene expression (18). The minimum P value method is used to group patients for survival analysis. This method does not require continuous biological knowledge nor assumptions to determine the best entry point for the continuous gene expression measurement. Empirical grouping of median/quartile/quartile does not necessarily reflect the true biological function of a gene. Therefore, the minimum P value method is a full-scale analysis method to explore the best critical point of risk separation for continuous variables. It has been shown to be highly practical in tumor size analysis, cell cycle measurements, and gene copy number analysis. Herein, we assessed the association between *DNM1L* and the outcomes of LUAD patients.

Kaplan-Meier plotter

The Kaplan-Meier plotter database (https://kmplot.com/analysis/index.php?p=service&cancer=pancancer_rnaseq) is available for evaluating the correlation of gene expression [microRNA (miRNA), messenger RNA (mRNA), protein]

with survival in more than thirty thousand samples of 21 tumor types from the TCGA and Gene Expression Omnibus (GEO) databases. This tool aims to conduct a meta-analysis-based identification and validation of potential survival biomarkers (19,20). We categorize patients based on the automatically selected optimal threshold. When this checkbox is ticked, we calculate all potential threshold values between the lower and upper quartiles of *DNMIL* expression, and then use the most effective threshold as the cutoff. The results page will show the false discovery rate along with the P value. Herein, the database was applied to evaluate the association between *DNMIL* expression and patient outcome, and to analyze the effect of both clinical variables and *DNMIL* on the survival outcome of LUAD patients.

TIMER2.0

The TIMER2.0 database (<https://cistrome.shinyapps.io/timer/>) provides a platform to comprehensively analyze the immune infiltrates in cancers. The extent of infiltration of immune cell types were evaluated using the TIMER algorithm. This allows researchers to analyze immune cell infiltration in tumor tissues with RNA sequencing (RNA-seq) profiles (21). Using the TIMER “gene” module, we explored the correlation between immune cell infiltration and *DNMIL* expression in LUAD. After inputting gene symbol for *DNMIL*, the cancer type for LUAD and the cell type of immune infiltration, we clicked the submit button to visualize a scatter plot, with illustrates the relationship between the estimated infiltrates value and gene expression. We defined the correlation between immune cell infiltration and *DNMIL* expression based on the following criteria: positive correlation ($P < 0.05$; Pearson correlation coefficient > 0); negative correlation ($P < 0.05$; Pearson correlation coefficient < 0). In addition, we also evaluated the correlation between *DNMIL* expression and hub gene sets of different immune cells in the TIMER2.0 database.

Search Tool for the Retrieval of Interacting Genes/Proteins (STRING) analysis

The STRING database (<https://cn.string-db.org/>) can be used to retrieve known and predicted protein-protein interactions (PPIs). In addition to producing exquisite networks of proteins, it also provides analysis of input proteins, including common functional enrichment analyses

and reference publications (22). The goal of STRING is to achieve a full interaction network, including physical interactions (direct interactions) and functional associations (indirect interactions).

GeneMANIA

GeneMANIA (<https://genemania.org/>) helps the users estimate the value of the target genes or gene sets and to identify additional genes that may have been missed in the screening process or novel genes with specific biological functions based on the function-related data. These include gene-protein interactions, molecular pathways, co-localization, co-expression, and protein domain similarity (23). In our study, the GeneMANIA database was used to develop the gene-gene interaction network involved in *DNMIL* to explore the potential molecular functions of the genes.

Cell lines

The human LUAD cell lines, including H1299, PC-9, A549, HCC827, H460, and H1975, were purchased from the China Center for Type Culture Collection (CCTCC, Wuhan, China) and were routinely cultured in RPMI 1640 containing 10% fetal bovine serum (FBS; Corning, Manassas, VA, USA) and 1% penicillin-streptomycin liquid (Solarbio, Beijing, China) at 37 °C with 5% CO₂.

RNA interference

Small interference RNAs (siRNAs) were purchased from RiboBio (Guangzhou, China) and used for *DNMIL* knockdown in LUAD cells. Three siRNAs were transiently transfected into cells by lipofectamine 2000 (Invitrogen, Carlsbad, CA, USA) at a concentration of 100 nM. The siRNA sequences: (I) 5'-3': GGAACAAAGTATCTTGCTA; (II) 5'-3': GAAGGGTTATTCCAGTCAA; and (III) 5'-3': GGTTGGAGATGGTGTTCAA.

Reverse transcription-quantitative polymerase chain reaction (RT-qPCR)

Total RNA was extracted by FastPure Cell/Tissue Total RNA Isolation Kit V2 (Vazyme, Nanjing, China) according to the manufacturer's instructions. Reverse transcription was performed with 1 µg of total RNA by HiScript III RT

SuperMix for qPCR (Vazyme). A Taq Pro Universal SYBR qPCR Master Mix (Vazyme) was selected for quantification. The relative gene expression was determined by the $2^{-\Delta\Delta C_t}$ method. The gene PCR primers were as follows: *β -actin* forward: TGGCACCCAGCACAATGAA, *β -actin* reverse: CTAAGTCATAGTCCGCCTAGAAGCA; *DNM1L* forward: GCAACTGGAGAGGAATGCTG, *DNM1L* reverse: GCACATCTAGCAGGTTACAG.

Western blot

Cells were lysed in RIPA Lysis Buffer (Beyotime Biotechnology, Shanghai, China) and proteins were separated by 4–12% gradient polyacrylamide gel electrophoresis (PAGE; YAMAY BIOTECH, Shanghai, China). Subsequently, the proteins were transferred to a PVDF membrane (Millipore Corp., Bedford, MA, USA), and blocked with 5% skimmed milk at room temperature for 2 h. Membranes were washed for three times with TBST (tris-buffered saline Tween-20) and were subsequently immunoblotted with the primary antibodies at 4 °C overnight. Then, the secondary antibodies (Beyotime Biotechnology) were incubated with these membranes at room temperature for 2 h. Finally, the immunoreactive bands were developed using an ECL detection system (BLT, PHOTO Technology, Guangzhou, China). The primary antibodies were as follows: *β -actin* (1:1,000, Abways, Shanghai, China), *DNM1L* (1:6,000, Proteintech, Wuhan, China).

Cell proliferation assays

Cell growth was detected by the cell counting kit-8 (CCK-8) assays (Dojindo, Kumamoto-ken, Japan). Transiently transfected H1299 and PC-9 cells with *DNM1L* siRNAs were seeded in 96-well plates with a density of 3,000 cells/well. Subsequently, CCK-8 was added to 96-well plates at 0, 24, 48, 72 h and incubated for another 2 h at 37 °C. A microplate reader (BioTek, Burlington, VT, USA) was used to measure the absorbance at 450 nm.

Colony formation assays

Transiently transfected H1299 and PC-9 cells with *DNM1L* siRNAs were seeded in 6-well plates at a density of 300 cells/well and incubated for 7 days. Then, the cells were fixed in paraformaldehyde for 30 minutes and stained with 0.1% crystal violet for 30 minutes.

Wound healing assays

Transiently transfected H1299 and PC-9 cells with siRNAs were seeded in 24-well plates at a density of 2×10^5 cells/well. A 10 μ L tip was used to scribe vertically on the bottom surface of 24-well plates. Subsequently, the cells underwent two rounds of phosphate buffer solution (PBS) washing followed by the addition of serum-free medium. Photographic documentation was conducted at 0, 24, and 48 h, respectively. The width measurements were calculated using Image J software.

Transwell assays

Transiently transfected H1299 and PC-9 cells with siRNAs were re-suspended in serum-free medium and seeded 5×10^4 cells with 200 μ L in the upper chamber (Corning, Manassas, VA, USA). For the lower chamber, medium containing 20% FBS was added. After 48 h, the cells in the upper chamber were discarded, while the cells crossing the chamber were fixed with paraformaldehyde and stained with crystal violet.

Statistical analysis

The GEPIA, Kaplan-Meier plotter, PrognScan, and TIMER2.0 databases were used to generate survival plots. STRING and GeneMANIA databases were used to develop the gene-gene interaction network. Spearman correlation analysis was used to determine the correlations between cohorts. Student *t*-test was used to determine the discrepancy between groups. For *in vitro* experiments, each experiment was repeated three times. Data were considered significant where P values were ≤ 0.05 (*, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$).

Results

DNM1L is increased in cancer

The expression of *DNM1L* in cancer and healthy tissues was evaluated using the TIMER2.0 database. The results indicated that *DNM1L* was significantly upregulated in tumor tissues compared to normal tissues in almost all the cancer datasets examined, including breast cancer (BRCA), cholangiocarcinoma (CHOL), esophageal carcinoma (ESCA), colon adenocarcinoma (COAD), glioblastoma (GBM), kidney renal clear cell carcinoma (KIRC), head and neck squamous cell carcinoma (HNSCC), LUAD,

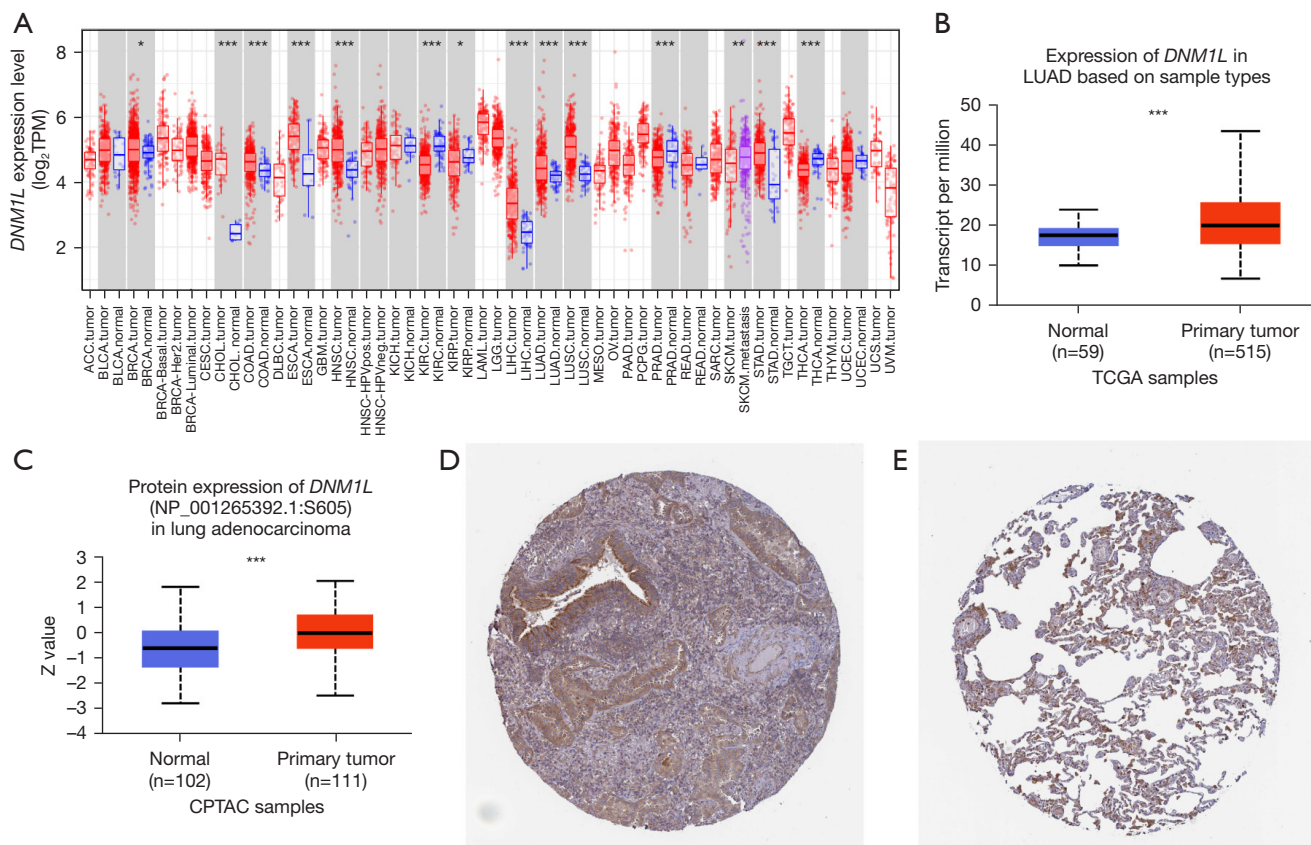


Figure 1 *DNMI1* expression in LUAD. (A) *DNMI1* mRNA expression in different cancer types based on the TIMER2.0 database. (B) *DNMI1* mRNA expression in LUAD tissues relative to normal lung tissues (UALCAN). (C) Expression levels of *DNMI1* phosphoprotein (CPTAC). Representative IHC staining of *DNMI1* protein expression in (D) LUAD tissue (<https://www.proteinatlas.org/ENSG00000087470-DNM1L/pathology/lung+cancer#img>) (quantity >75%) and (E) normal lung tissue (<https://www.proteinatlas.org/ENSG00000087470-DNM1L/tissue/lung#img>) (quantity 25–75%) in the HPA database. *, P<0.05; **, P<0.01; ***, P<0.001. *DNMI1*, dynamin 1-like; TPM, transcript per million; LUAD, lung adenocarcinoma; TCGA, The Cancer Genome Atlas; CPTAC, Clinical Proteomic Tumor Analysis Consortium; mRNA, messenger RNA; TIMER, Tumor IMMune Estimation Resource; UALCAN, University of The Alabama at Birmingham CANcer data analysis Portal; IHC, immunohistochemistry; HPA, Human Protein Atlas.

liver hepatocellular carcinoma (LIHC), lung squamous cell carcinoma (LUSC), thyroid cancer (THCA), and stomach adenocarcinoma (STAD) (P<0.05) (Figure 1A). In addition, we estimated *DNMI1* mRNA expression between the LUAD and healthy tissues based on the TCGA data in the UALCAN database. Our results demonstrated that *DNMI1* mRNA was significantly upregulated in LUAD (P<0.05), which is consistent with data from the TIMER2.0 database (Figure 1B). Moreover, phosphoprotein levels of *DNMI1* in tumor tissue were significantly higher than that in the healthy tissue based on CPTAC data from the UALCAN database (Figure 1C). Similarly, IHC staining data in the HPA database indicate that *DNMI1* protein expression

is increased in LUAD tissues compared to healthy lung tissues (P<0.05) (Figure 1D,1E). These results demonstrated that *DNMI1* is upregulated in LUAD, suggesting that *DNMI1* may play essential roles in the tumorigenesis and progression.

***DNMI1* expression correlates with clinical and molecular parameters**

To further explore the correlation of *DNMI1* expression with clinical variables, we analyzed the mRNA profiles of *DNMI1* in subgroups of patients with different clinical variables based on UALCAN database analysis. Data

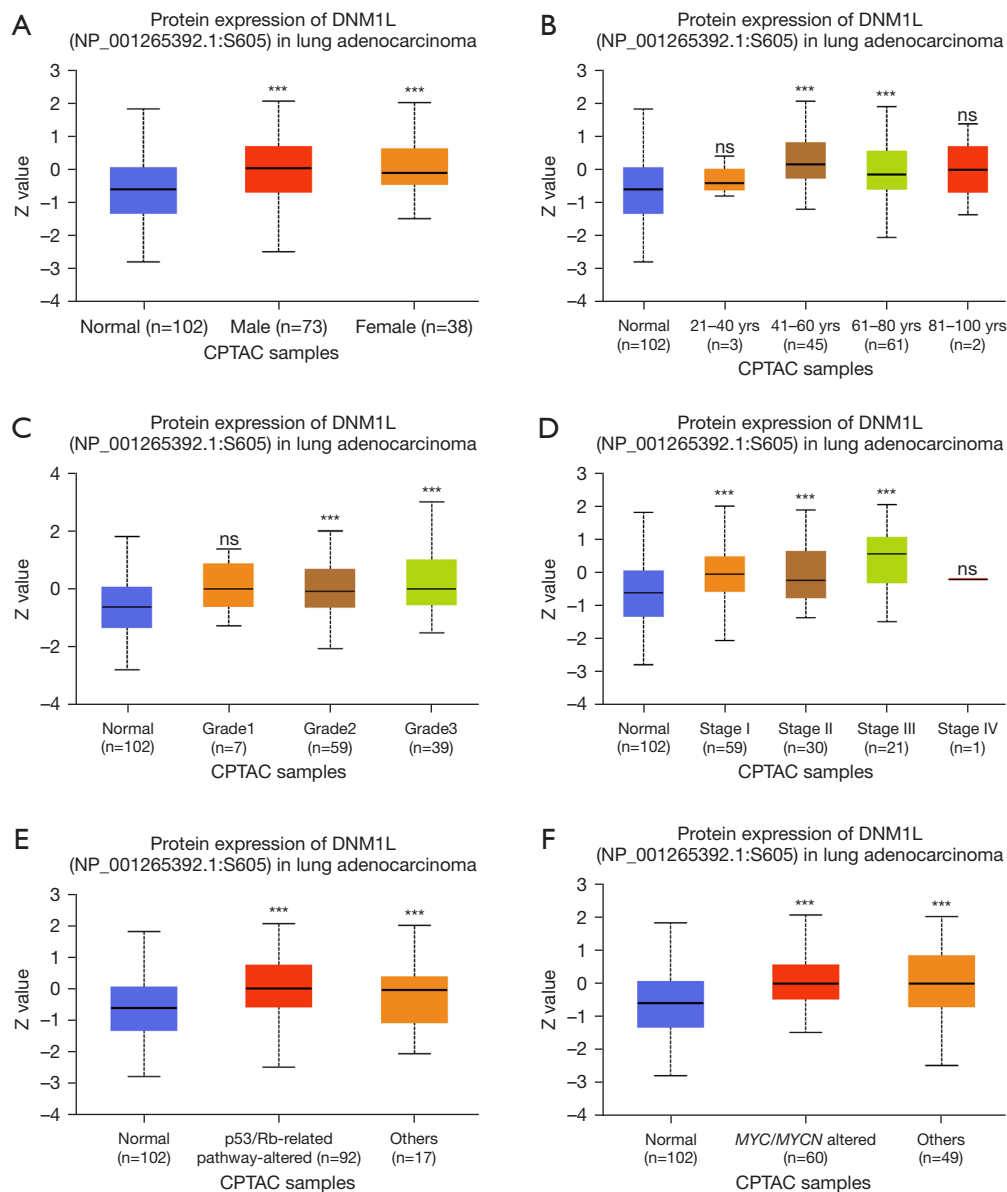


Figure 2 Correlation between DNMI1 protein expression and clinical and molecular parameters including (A) gender, (B) age, (C) tumor grade (grade 1, 2, or 3), (D) tumor stage (stage I, II, III, or IV), (E) p53/Rb status (p53/Rb-related pathway-altered LUAD tumors and others: p53 WT LUAD tumors), and (F) *MYC/MYCIN* status (*MYC/MYCIN* altered LUAD tumors and others: *MYC/MYCIN* WT LUAD tumors) in LUAD patients based on CPTAC data analysis. ***, $P < 0.001$; ns, not significant. DNMI1, dynamin 1-like; CPTAC, Clinical Proteomic Tumor Analysis Consortium; *MYC*, myelocytomatosis oncogene; *MYCIN*, neuroblastoma-derived *MYC*; *TP53*, tumor protein p53; LUAD, lung adenocarcinoma; WT, wild type.

showed that DNMI1 protein expression was significantly upregulated in both female and male LUAD tissue relative to adjacent normal lung tissue ($P < 0.05$) (Figure 2A). Additionally, DNMI1 was significantly upregulated in LUAD tissue in patients across different age groups: 41–60 and 61–80 years, compared to DNMI1 expression

in adjacent normal lung tissues ($P < 0.05$) (Figure 2B). Significantly higher expression of DNMI1 was observed in grade 2 and grade 3 LUAD tumors in addition to patients with stage 1, 2, and 3 diseases ($P < 0.05$) (Figure 2C, 2D). There was also a significantly higher expression of DNMI1 protein between grade 3 and grade 2 tumors.

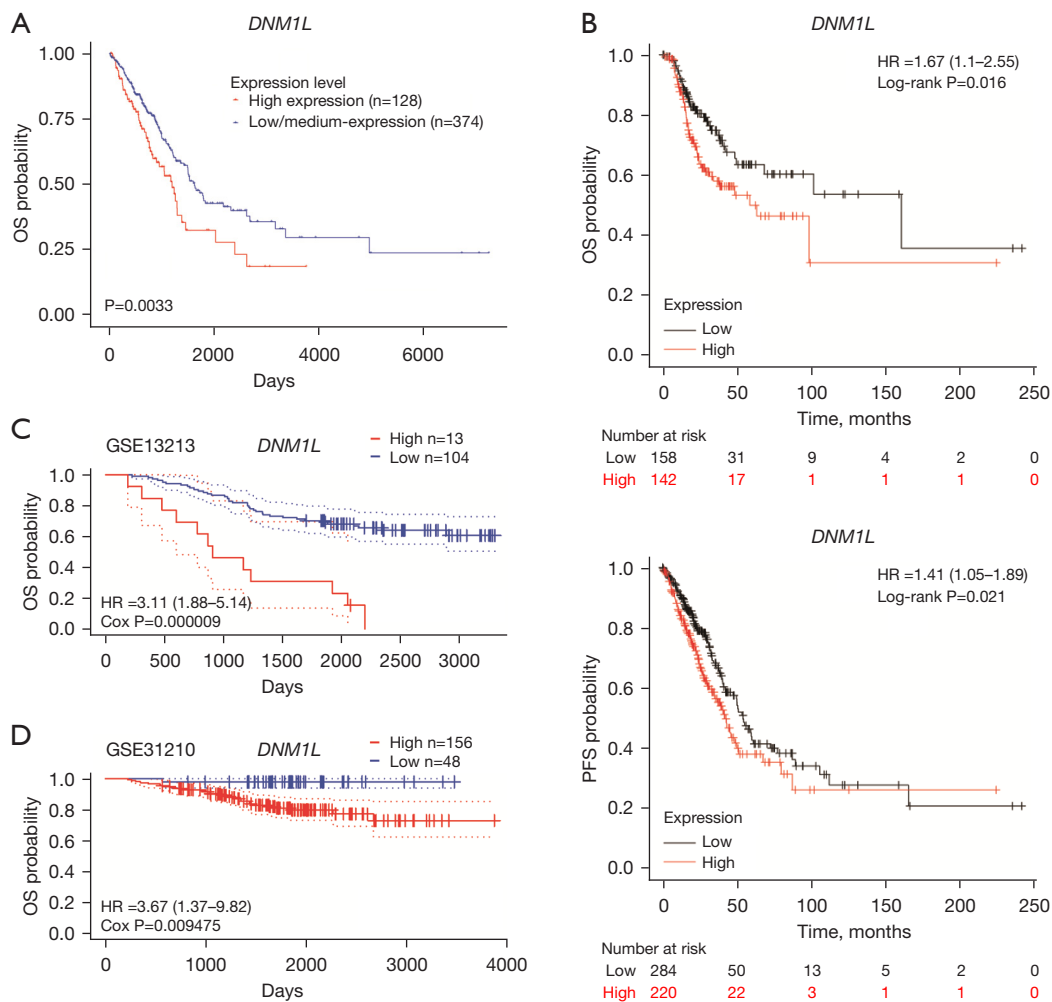


Figure 3 Correlation between *DNMI1* expression and outcomes of LUAD patients. (A) High expression of *DNMI1* is associated with unfavorable outcomes (UALCAN). (B) High *DNMI1* expression is associated with shorter OS and PFS of LUAD patients (Kaplan-Meier plotter). *DNMI1* expression also correlates with survival outcomes in patients from the (C) GSE13213 and (D) GSE31210 datasets (PrognScan). *DNMI1*, dynamin 1-like; OS, overall survival; HR, hazard ratio; PFS, progression-free survival; LUAD, lung adenocarcinoma; UALCAN, University of The ALabama at Birmingham CANcer data analysis Portal.

DNMI1 expression was significantly higher in p53/Rb-related pathway-altered cohorts compared to p53 wild-type (WT) LUAD tumors and adjacent normal lung tissue ($P < 0.05$) (Figure 2E). Furthermore, *DNMI1* expression was significantly upregulated in both myelocytomatosis oncogene (*MYC*)/neuroblastoma-derived *MYC* (*MYCN*) altered and other (*MYC*/*MYCN* WT) LUAD samples compared to adjacent normal lung tissue ($P < 0.05$) (Figure 2F). Overall, these data highlight significant increases in *DNMI1* protein expression across a variety of clinical and molecular variables in LUAD patients.

DNMI1 is an independent prognostic biomarker in LUAD

The correlation between *DNMI1* expression and prognosis of LUAD patients was evaluated using the UALCAN database. Kaplan-Meier survival curves showed that high expression of *DNMI1* was significantly associated with unfavorable outcomes in patients with LUAD ($P = 0.0033$) (Figure 3A). Patients with high expression of *DNMI1* had significantly worse overall survival (OS) and progression-free survival (PFS) compared to those with low *DNMI1*

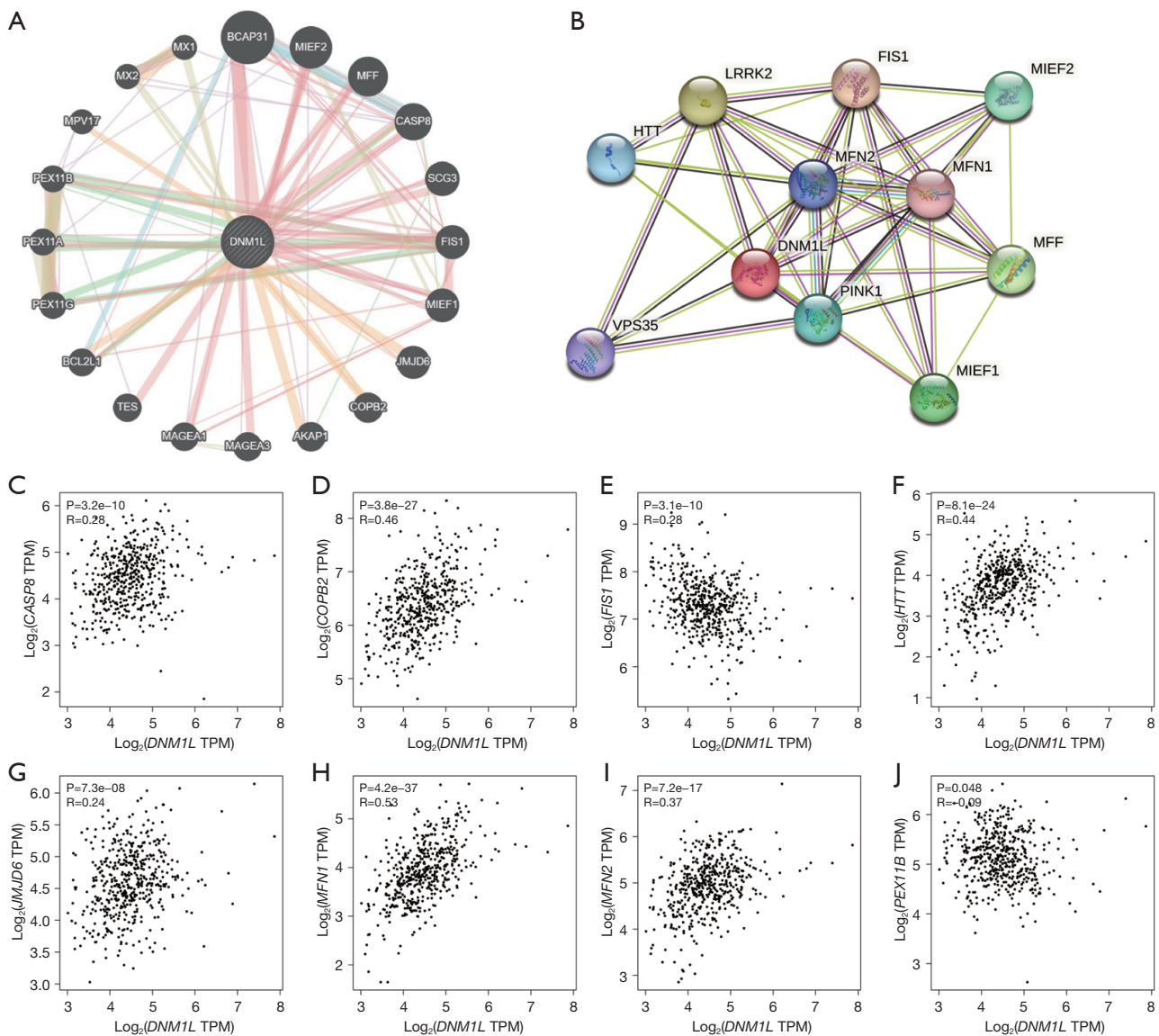


Figure 4 PPI network analysis. (A) The *DNM1L* and *DNM1L* gene network was analyzed using the GeneMANIA database. (B) PPIs between *DNM1L* protein and *DNM1L*-related proteins were determined using STRING. (C-J) Spearman correlation analysis showing the correlations between *DNM1L* expression and *CASP8*, *COPB2*, *FIS1*, *HTT*, *JMJD6*, *MFN1*, *MFN2*, and *PEX11B* in LUAD are shown. *DNM1L*, dynamin 1-like; TPM, transcript per million; PPI, protein-protein interaction; STRING, Search Tool for the Retrieval of Interacting Genes/Protein; LUAD, lung adenocarcinoma.

expression, based on microarray data of *DNM1L* expression from the Kaplan-Meier plotter database ($P < 0.05$) (Figure 3B). Furthermore, we validated the predictive value in the independent GEO datasets from the PrognScan database. Consistent with the above data, these results demonstrated that high expression of *DNM1L* was significantly associated with worse OS in two independent GEO cohorts (GSE13213 and GSE31210) ($P < 0.05$) (Figure 3C, 3D).

DNM1L interacts with genes involved in mitochondrial, apoptotic, and peroxisomal function

Gene-gene interactions were constructed to evaluate genes that interact with *DNM1L* in LUAD and to explore the functions of these genes using the GeneMANIA database. A total of 20 genes were significantly associated with *DNM1L* (Figure 4A). The top 5 candidate genes that

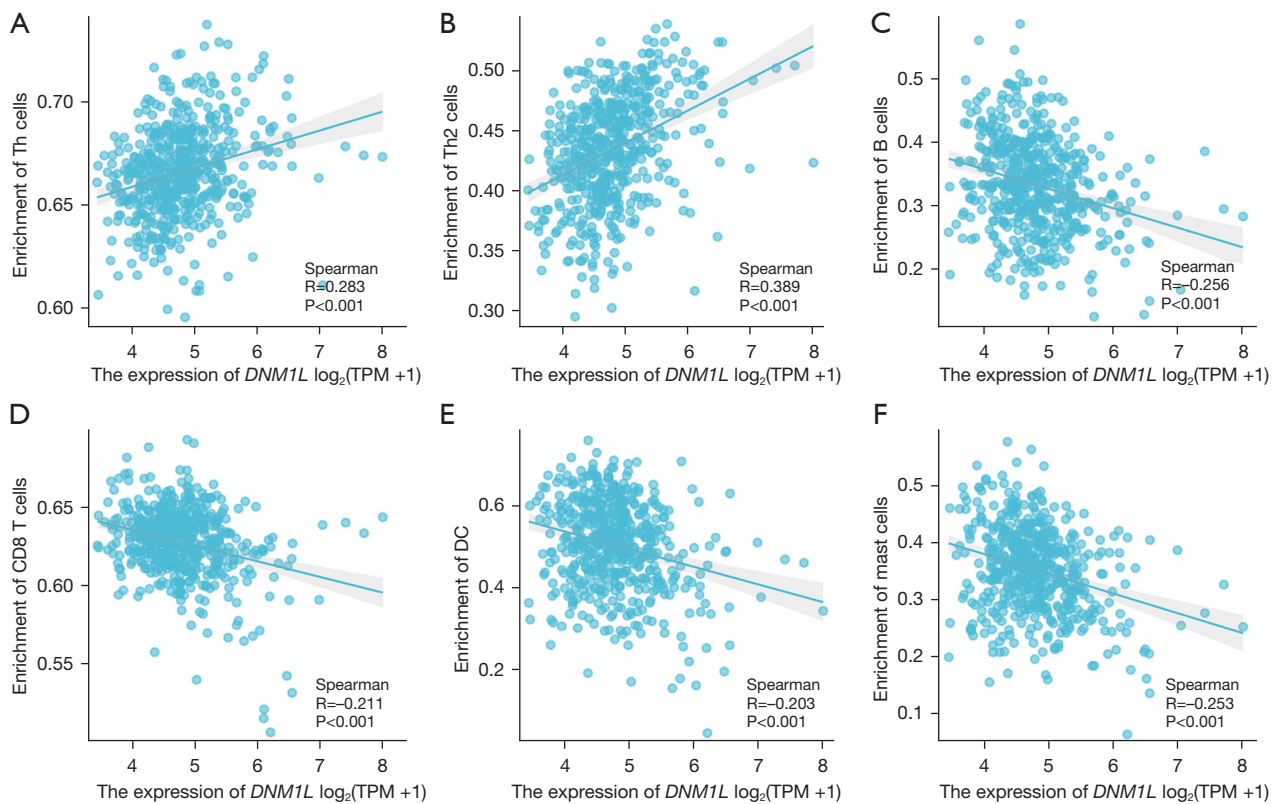


Figure 5 Correlation between *DNM1L* expression and infiltration levels of immune cell subsets. (A) Th cells, (B) Th2 cells, (C) B cells, (D) CD8 T cells, (E) DCs, and (F) mast cells. *DNM1L*, dynamin 1-like; Th, T helper; TPM, transcript per million; DC, dendritic cell.

correlated with *DNM1L* gene expression included B cell receptor-associated protein 31 (*BCAP31*), mitochondrial elongation factor 2 (*MIEF2*), mitochondrial fission factor (*MFF*), caspase-8 (*CASP8*), and secretogranin III (*SCG3*). Furthermore, functional analysis on the differentially expressed genes (DEGs) between the high and low *DNM1L* expression cohorts demonstrated that these genes were mainly associated with the following biological processes; microbody, peroxisome organization, regulation of mitochondrial fission, mitochondrial fission, and regulation of mitochondrion organization. To further evaluate the functional mechanisms of the *DNM1L* gene, its associated PPI network was examined using STRING. The top 11 proteins identified in the *DNM1L* network were shown (Figure 4B). Among these, the five proteins with the highest degree of centrality were mitochondrial fission 1 (*FIS1*), huntingtin (*HTT*), leucine-rich repeat kinase 2 (*LRRK2*), *MFF*, and *MIEF2*. In addition, we further estimated the association between *DNM1L* and the interacting genes in the GEPIA database. Further

experiments found that the expression of *MFF* in LUAD was increased, while the expression of *FIS1* was decreased (Figure S1). Our results demonstrated that *DNM1L* was significantly associated with *CASP8*, *COPB2*, *FIS1*, *HTT*, *JMJ3D6*, *MFN1*, *MFN2*, and *PEX11B* in LUAD ($P < 0.05$) (Figure 4C-4J).

DNM1L expression correlates with immune cell infiltration in LUAD

Lymphocyte infiltration is known to be an independent factor of survival outcomes and sentinel lymph node status. Herein, we explored the correlation between *DNM1L* expression and the immune cell infiltration in the TME. We performed the Spearman correlation analysis to evaluate the correlation between *DNM1L* expression and infiltration levels of immune cell subsets based on the data from TCGA-LUAD cohort using ggplot2 R package in R software (4.2.1 version). As shown in Figure 5, the results showed that *DNM1L* expression was significantly positively

Table 1 Correlation between *DNM1L* and immune cell markers (TIMER2.0)

Immune cell	Gene marker	Correlation	P value
CD8 ⁺ T cell	<i>CD8A</i>	0.103	*
	<i>CD8B</i>	0.091	*
T cell (general)	<i>CD3D</i>	0.005	0.910
	<i>CD3E</i>	0.025	0.577
	<i>CD2</i>	0.006	0.892
B cell	<i>CD19</i>	0.004	0.921
	<i>CD79A</i>	0.002	0.956
Monocyte	<i>CD86</i>	0.116	**
	<i>CSF1R</i>	0.080	0.069
TAM	<i>CCL2</i>	0.123	**
	<i>CD68</i>	0.105	*
	<i>IL10</i>	0.116	**
M1	<i>IRF5</i>	0.048	0.275
	<i>NOS2</i>	0.212	***
	<i>PTGS2</i>	0.198	***
M2	<i>CD163</i>	0.247	***
	<i>VSIG4</i>	0.074	0.094
	<i>MS4A4A</i>	0.082	0.062
Neutrophils	<i>CEACAM8</i>	-0.127	**
	<i>ITGAM</i>	0.063	0.155
	<i>CCR7</i>	-0.010	0.828
NK cell	<i>KIR2DL1</i>	0.062	0.160
	<i>KIR2DL3</i>	0.195	***
	<i>KIR2DL4</i>	0.238	***
	<i>KIR3DL1</i>	0.141	**
	<i>KIR3DL2</i>	0.161	***
	<i>KIR2DS4</i>	0.091	*
Dendritic cell	<i>HLA-DPB1</i>	-0.264	***
	<i>HLA-DQB1</i>	-0.229	***
	<i>HLA-DRA</i>	-0.235	***
	<i>HLA-DPA1</i>	-0.194	***
	<i>CD1C</i>	-0.297	***
	<i>NRP1</i>	0.151	***
	<i>ITGAX</i>	0.103	*

Spearman correlation analysis was used to determine correlations, with statistical significance represented as: *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. *DNM1L*, dynamin 1-like; TIMER, Tumor IMMune Estimation Resource; TAM, tumor-associated macrophage; NK, natural killer.

associated with the infiltration levels of T helper (Th) cells, especially Th2 cells. while, negatively associated with the infiltration levels of B cells, CD8 T cells, dendritic cells, and mast cells ($P < 0.05$).

Correlation between immune marker sets and DNM1L expression

We further evaluated the correlation between infiltrating immune cell subtypes and *DNM1L*, based on immune cell markers using TIMER2.0. Our data showed that *DNM1L* was significantly associated with numerous immune cell markers ($P < 0.05$) (Table 1). Subsequently, we explored the association of *DNM1L* expression with diverse populations of functional T cells, such as regulatory T cells (Tregs), effector T cells, Th1, Th2, memory, naïve, and exhausted T cells using the TIMER2.0 database. Of interest, *DNM1L* expression was significantly associated with 28 of the 41 T cell markers in LUAD (Table 2), further supporting our data that *DNM1L* expression is positively related to the infiltration of anti-tumor immune cells.

DNM1L promotes the proliferation and migration of LUAD cells

The six LUAD cell lines (H1299, PC-9, A549, HCC827, H460 and H1975) selected are all commonly used in lung cancer research. For validation of our predicted tumor promoter role of *DNM1L*, we first detected the RNA and protein expression of *DNM1L* in different LUAD cell lines (Figure 6A-6C), and selected two cell lines H1299 and PC-9 with relatively high expression of *DNM1L* for further experiments. Then, we used siRNAs to inhibit the expression of *DNM1L* in H1299 and PC-9 cells. In the pre-experiment, all three siRNAs can effectively knock down the *DNM1L* (Figure 6D,6E), so we use the mixture of three siRNAs to construct the *DNM1L*-knockdown cell lines, and the silence of *DNM1L* expression was confirmed by RT-qPCR and western blot (Figure 6F-6I).

The CCK-8 assays and colony formation assays demonstrated that the knockdown of *DNM1L* significantly suppressed the proliferation ability of LUAD cells (Figure 7A-7E). Meanwhile, the results of wound healing assays and transwell assays showed that the knockdown of *DNM1L* expression remarkably decreased the migration capacity of LUAD cells (Figure 7F-7J). These results confirmed that *DNM1L* is a detrimental factor in LUAD.

Table 2 Correlation between *DNM1L* and markers associated with specific T cell subsets (TIMER2.0)

T cell	Gene marker	Correlation	P value
Th1	<i>STAT4</i>	-0.015	0.742
	<i>STAT1</i>	0.346	***
	<i>IFNG</i>	0.163	***
	<i>TNF</i>	0.094	*
Th1-like	<i>CXCL13</i>	0.034	0.280
	<i>HAVCR2</i>	0.075	*
	<i>BHLHE40</i>	0.105	*
	<i>CD4</i>	0.061	0.166
	<i>IL21</i>	0.179	***
Th17	<i>STAT3</i>	0.267	***
	<i>IL17A</i>	0.067	0.128
Treg	<i>FOXP3</i>	0.138	***
	<i>CCR8</i>	0.212	***
	<i>STAT5B</i>	0.287	***
	<i>TGFB1</i>	-0.035	0.424
Effector T cell	<i>FGFBP2</i>	-0.043	0.331
	<i>FCGR3A</i>	0.193	***
Naïve T cell	<i>SELL</i>	0.038	0.219
	<i>TCF7</i>	0.196	***
Effective memory T cell	<i>PDCD1</i>	0.090	*
	<i>DUSP4</i>	0.350	***
	<i>GZMK</i>	0.032	0.470
	<i>GZMA</i>	0.101	*
	<i>IFNG</i>	0.163	***
Resistant memory T cell	<i>CD69</i>	-0.049	0.267
	<i>ITGAE</i>	0.057	0.200
	<i>CXCR6</i>	0.095	*
	<i>MYADM</i>	0.179	***
Exhausted T cell	<i>HAVCR2</i>	0.075	*
	<i>TIGIT</i>	0.172	***
	<i>LAYN</i>	0.107	*
Resting Treg T cell	<i>FOXP3</i>	0.138	***
	<i>IL2RA</i>	0.291	***

Table 2 (continued)

Table 2 (continued)

T cell	Gene marker	Correlation	P value
Effective Treg T cell	<i>FOXP3</i>	0.138	***
	<i>CTLA4</i>	0.132	*
	<i>CCR8</i>	0.212	***
General memory T cell	<i>TNFRSF9</i>	0.172	***
	<i>SELL</i>	0.038	0.219
	<i>IL7R</i>	0.199	***

Spearman correlation analysis was used to determine correlations, with statistical significance represented as: *, P<0.05; ***, P<0.001. *DNM1L*, dynamin 1-like; TIMER, Tumor Immune Estimation Resource; Th, T helper; Treg, regulatory T cell.

Discussion

The relevant prognostic and therapeutic markers play an important role in the management of LUAD patients (24,25). Dysregulated mitochondrial dynamics are responsible for the progression and metastasis of many cancers. The functional mitochondrial network is regulated by various key proteins, including *DNM1L*. These mitochondria-shaping proteins play important roles in diverse biological pathways such as cell metabolism, proliferation, migration, cell death, as well as innate and adaptive immunity via its regulation of the dynamic equilibrium between mitochondria fission and fusion. Notably, mitochondrial dynamics are altered, thereby resulting in the inhibition of apoptosis and altered metabolic requirements of cancer cells which help drive tumor initiation and progression.

Accumulating evidence indicates that *DNM1L*-induced mitochondrial fission acts as an important mechanism in carcinogenesis. Overexpression of *DNM1L* contributes to disease progression and induces glycolysis through regulation of the *FOXM1/MMP12* axis in head and neck cancer (26). Increased *DNM1L* also promotes autophagy and disease progression of esophageal squamous cell carcinoma (27). It has been reported that *DNM1L* may be recruited to the mitochondria to promote mitochondrial fission, facilitating cell metastasis in hepatocellular carcinoma cells both *in vivo* and *in vitro* (28). It can also promote metabolic changes induced by *KRAS* to drive the growth of pancreatic tumors (29). In addition, *DNM1L*-mediated mitochondrial fission participates in the development of

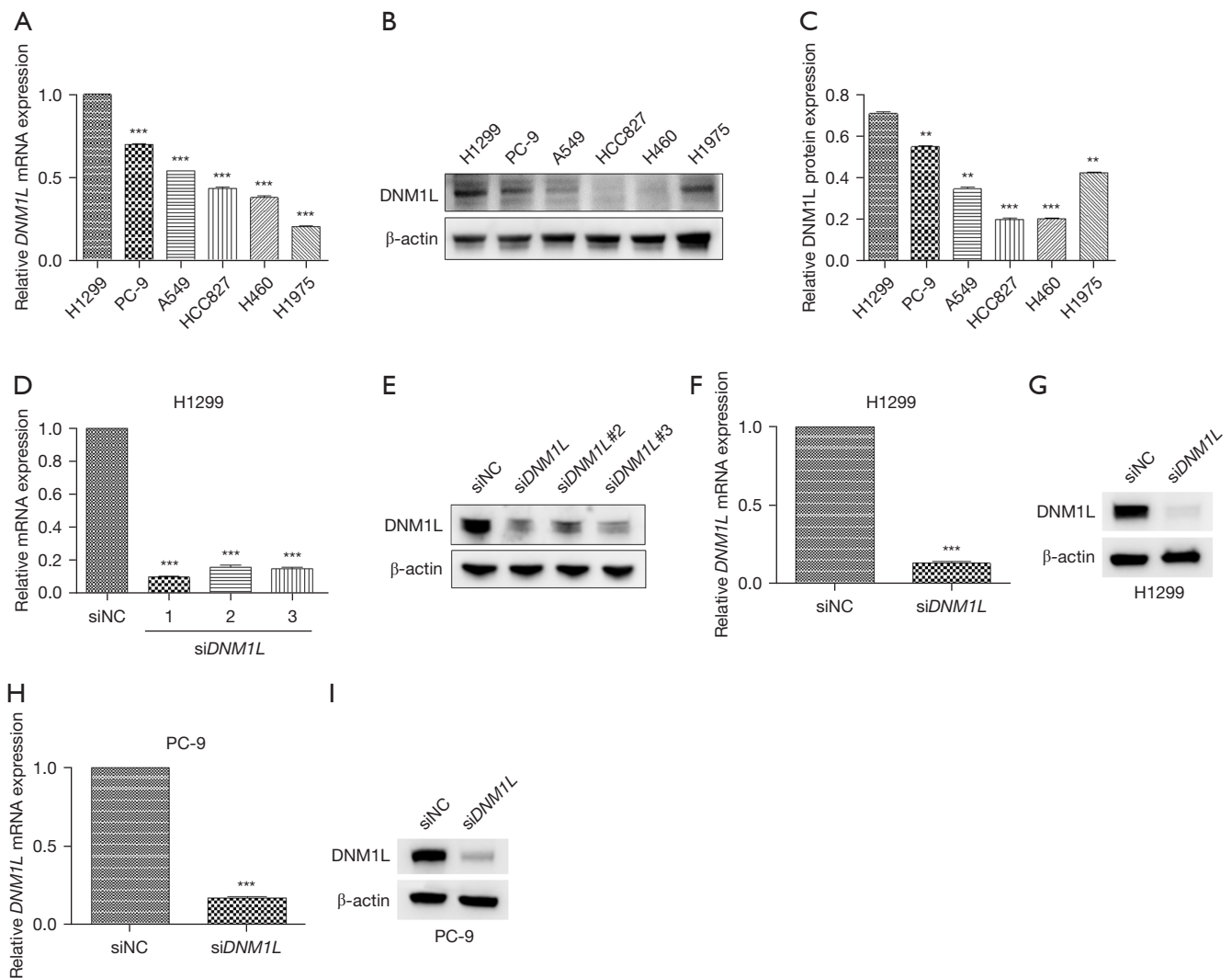


Figure 6 Construction of *DNM1L* knockdown cell lines. (A) The relative mRNA expression of *DNM1L*. (B,C) The relative protein expression of *DNM1L*. (D,E) Knocking efficiency of siRNAs of *DNM1L*. (F,G) The expression of *DNM1L* in constructed H1299 cell lines. (H,I) The expression of *DNM1L* in constructed PC-9 cell lines. **, $P < 0.01$; ***, $P < 0.001$. *DNM1L*, dynamin 1-like; mRNA, messenger RNA; si, small interfering; siNC, siRNA negative control.

chemoresistance in various cancer types (30). *DNM1L*- and *MIEF1/2*-mediated mitochondrial fission has been shown to link the extracellular matrix mechano-transduction to metabolism and resistance to chemotherapy in metastatic BRCA (31). These observations point to *DNM1L* as a potential therapeutic target for cancer treatment and as such, warrants further studies to elucidate its role in this disease.

In view of the important role of elevated *DNM1L* in stem cell maintenance, cell metabolism, proliferation, invasion, and metastasis, various *DNM1L* inhibitors

have been identified in numerous studies as a potential therapeutic strategy in cancer treatment. Mitochondrial division inhibitor 1 (mdivi-1) is an efficacious *DNM1L* protein inhibitor, which can inhibit cell proliferation by inducing mitochondrial fusion and changes in oxygen consumption (32). Fang *et al.* reported that mdivi-1 could induce spindle abnormalities and augment the cytotoxicity of taxol to improve its efficacy in BRCA (33). Furthermore, Wang *et al.* demonstrated that mdivi-1 could enhance death receptor-mediated apoptosis in human ovarian cancer cells (34). This study also showed that SUMOylated

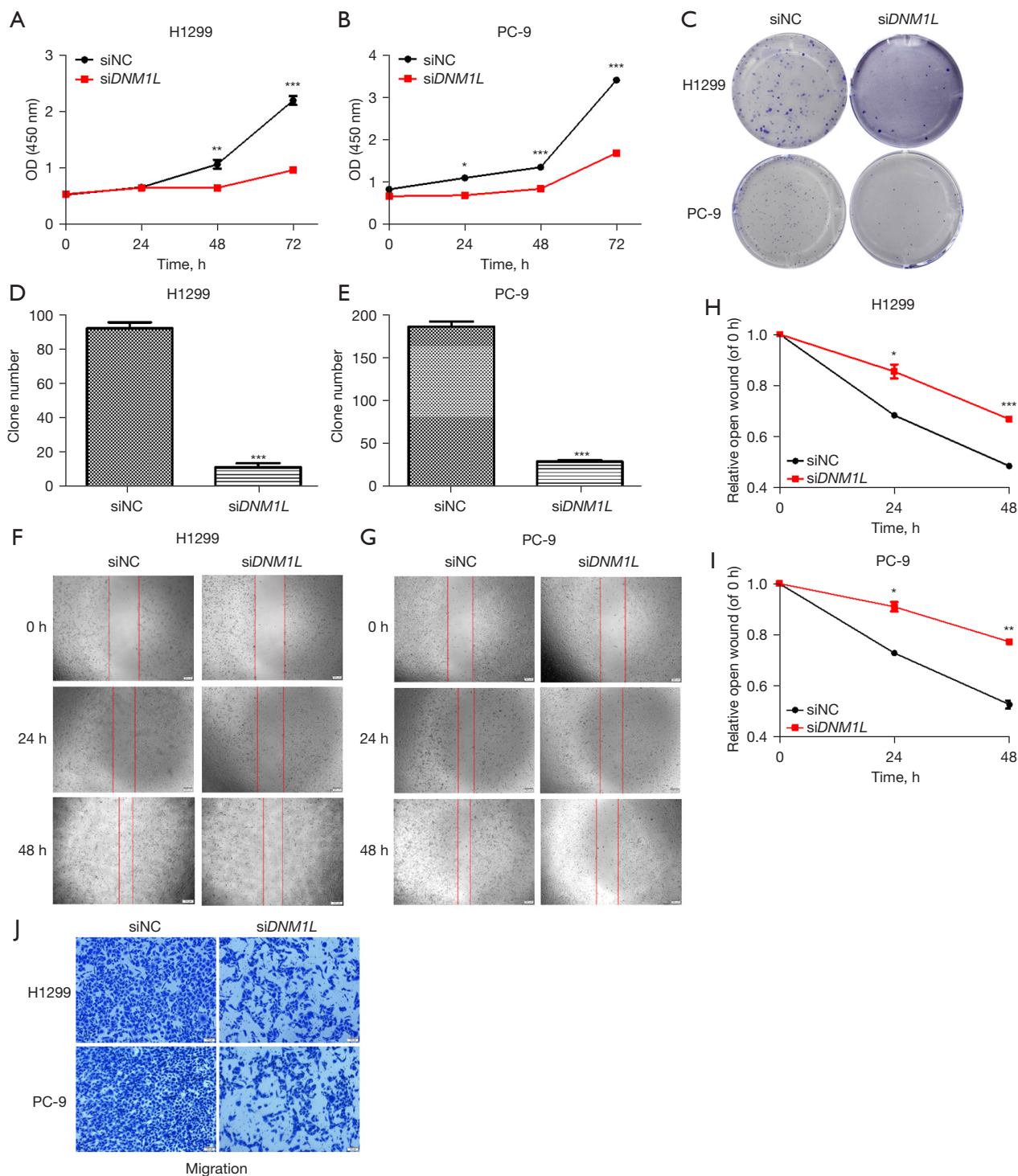


Figure 7 *DNM1L* promote the proliferation and migration of LUAD cells. (A,B) The proliferation ability of LUAD cells was detected by CCK-8 assays. (C-E) The proliferation ability of LUAD cells was detected by colony formation assays. Colonies were stained with crystal violet and imaged on the digital camera (Nikon, Japan). The colony numbers were counted using ImageJ. (F-J) The migration capacity of LUAD cells was detected by wound healing assays and transwell assays and stained with crystal violet. (F,G) Scale bar =200 μ m; (J) scale bar =100 μ m. *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. OD, optical density; si, small interfering; siNC, siRNA negative control; *DNM1L*, dynamin 1-like; LUAD, lung adenocarcinoma; CCK-8, cell counting kit-8.

synaptojanin 2 binding protein-cytochrome-c oxidase 16, recruits *DNM1L* to the mitochondria to enhance SUMOylation and phosphorylation in order to induce mitochondrial fission which could be inhibited by mdivi-1 (35). These findings suggest that treatments directed against the mitochondria via targeting of *DNM1L* protein, may represent a novel strategy for many diseases, including cancer. Despite this, the relationship between *DNM1L* and prognosis in LUAD remains largely unexplored, while a link between *DNM1L* expression and immune cell infiltration in the TME of this NSCLC subtype, remains to be elucidated. Further exploration of the expression patterns of *DNM1L* in LUAD and its relationship with specific immune cell subtypes is warranted.

In this study, we explored the clinical value and immune regulation of *DNM1L* in LUAD using data from the TCGA, UALCAN, and TIMER2.0 databases. Our data show that *DNM1L* is significantly overexpressed in LUAD. High *DNM1L* expression was associated with various clinical and molecular parameters, including higher tumor grade and stage, in addition to OS and PFS in patients with LUAD, thereby suggesting that *DNM1L* is an independent prognostic factor in LUAD. Immunotherapy using ICIs aims to enhance natural defenses to eliminate cancer cells. This has revolutionized anti-tumor treatments in the lung cancer setting and yielded unparalleled survival benefits for cancer patients. However, limited efficacy remains a challenge in most clinical settings and only a minority of patients benefit from this treatment, with even fewer achieving durable responses (36). The infiltrating immune cells in the TME are significantly associated with disease progression and therapeutic outcome of cancer patients (37). The TME contains a variety of cell types, including immune cells, stromal cells, and cancer cells as well as secreted factors which are often the targets for anti-tumor therapies (38). Advanced therapeutic strategies targeting the TME in cancer have been developed in recent years (39). Since tumor-infiltrating immune cells, in particular cytotoxic CD8⁺ T cells play essential roles in immunotherapy, comprehensive research of the immune cells within the TME in tumors is vital, in order to develop new immunotherapeutic strategies and to optimize treatments (40). Notably, the mitochondria are involved in inflammation, immunity and the TME in response to cellular stress, damage, and infection (41). Mitochondrial dynamics are crucial for many cellular processes such as mitochondrial integrity, apoptosis, cell cycle, inflammatory responses, and immunity (42,43). Mitochondrial dynamics

regulate the activation, differentiation, and cytokine production of immune cells, including B cells, T cells, dendritic cells, and macrophages, in both innate and adaptive immunity (14). Zheng *et al.* showed that excessive mitochondrial fission is associated with loss of NK cells and reduced cytotoxicity to evade the surveillance mediated by the NK cells, resulting in unfavorable survival in patients with liver cancer (12). The sustained activation of rapamycin-GTPase *DNM1L* of NK cells in the hypoxic TME results in excessive mitochondrial fission (12). PD-1 signaling downregulates *DNM1L* phosphorylation on serine 616 (Ser⁶¹⁶) to restrain mitochondrial fission mediated by T cells via the extracellular signal-regulated kinases 1 and 2 (*ERK1/2*) and mechanistic target of rapamycin (mTOR) pathways (44). Bao *et al.* demonstrated that overexpressing *DNM1L* was positively associated with infiltration of TAMs in the TME of hepatocellular carcinomas (45). *DNM1L*-mediated mitochondrial fission-induced cytosolic mitochondrial DNA (mtDNA) stress figure upregulates C-C motif chemokine 2 (CCL2) via toll-like receptor 9 (TLR9)-mediated nuclear factor- κ B (NF- κ B) signaling and thus, facilitates the polarization and recruitment of TAMs (45). As a systemic disease, cancer can regulate many functions and components of the immune system. Our data indicate that *DNM1L* participates in the tumor-immune interactions of LUAD and is associated with immune cell infiltration. We also report that *DNM1L* is positively associated with the level of infiltration of numerous immune cells in LUAD, including Th cells, B cells, CD8 T cells, dendritic cells, and mast cells. Furthermore, we show that *DNM1L* is significantly associated with diverse subtypes of T cells, including Th1, Th2, T follicular helper (Tfh), Treg, native, memory, effector, and exhausted T cells, suggesting that *DNM1L* plays a key role in immunity and therefore, may be a potential therapeutic target in the TME of LUAD patients.

Moreover, our study shows that tumor protein p53 (*TP53*) WT LUAD tumors had higher *DNM1L* expression compared to p53-mutated tumors, a finding of which is consistent with previous studies (46,47). Accumulating evidence has indicated that p53 can induce phosphorylation of *DNM1L* at Ser⁶³⁷ and subsequent mitochondrial dysfunction, thereby providing evidence for its important role in mitochondrial dynamics and function (48). Such findings support the idea that inhibition of *DNM1L* may be an effective therapeutic strategy for overcoming drug resistance in p53-mutated cancers, including LUAD (46,47).

Limitations

This study had a number of limitations. Firstly, the analyses were conducted based on data from several public databases. We are aware that web-based databases, including UALCAN, HPA, GEPIA, PrognoScan, Kaplan-Meier plotter, TIMER2.0, STRING, and GeneMANIA, lack detailed information about data sources and methodologies, which is a significant limitation in my study. Moreover, in the correlation analysis between *DNM1L* expression and clinical and molecular parameters, some groups have small sample size, even less than 10. Further research, both *in vivo* and *in vitro*, is required to examine the underlying molecular mechanisms of *DNM1L* in regulating the immune cell infiltration in LUAD. Secondly, future prospective studies with larger sample cohorts are warranted to evaluate the prognostic value of *DNM1L*. Thirdly, while mitochondrial fission is regulated by several genes, including *DNM1L*, *FIS1*, and *MFF* (49), this study focused primarily on the role of *DNM1L* which to our knowledge, has not been extensively reported in the context of LUAD.

Conclusions

Taken together, our findings demonstrate *DNM1L* as a potential prognostic biomarker in LUAD. Furthermore, we report for the first time in LUAD, a role for *DNM1L* in regulating proliferation and migration of LUAD cells as well as the infiltration of tumor-related immune cells, which suggests *DNM1L* was a potential therapeutic target in LUAD. Further studies are however warranted to define its exact mechanism of action and potential therapeutic significance in LUAD patients.

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Footnote

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Conflicts of Interest: All authors have completed the ICMJE uniform disclosure form (available at <https://tclr.amegroups.com/article/view/10.21037/tclr-23-685/coif>). The authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

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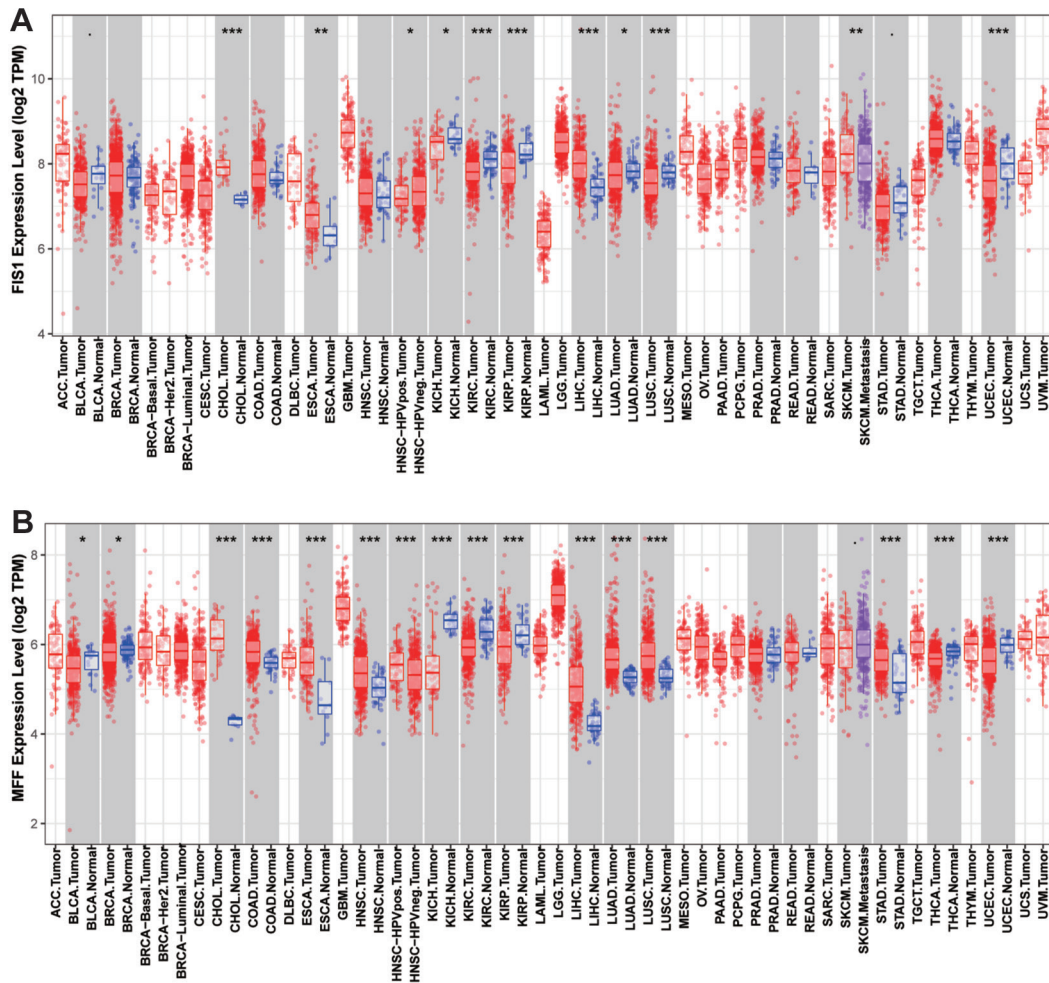


Figure S1 (A) *FIS1* and (B) *MFF* mRNA expression levels in different cancer types based on TIMER2.0. *, $P < 0.05$; **, $P < 0.01$; ***, $P < 0.001$. *FIS1*, mitochondrial fission 1; TPM, transcript per million; *MFF*, mitochondrial fission factor; mRNA, messenger RNA; TIMER, Tumor Immune Estimation Resource.