Diagnostic Values of METTL1-Related Genes and Immune Characteristics in Systemic Lupus Erythematosus

Yu Liu1,*, Enyi Zhu1,*, Yan Lei1,*, Ailing Luo2, Yaping Yan2, Mansi Cai2, Shanshan Liu2, Yan Huang3, Hui Guan1, Ming Zhong1, Weinian Li9, Lian Lin1, Michael Hultstöm5,6, Enyin Lai7, Zhihua Zheng1, Xiaoping Liu2, Chun Tang1

1Department of Nephrology, Center of Kidney and Urology, the Seventh Affiliated Hospital, Sun Yat-sen University, Shenzhen, 518107, People’s Republic of China; 2Department of Hematology, Guangzhou Women and Children’s Medical Center, Guangzhou Medical University, Guangdong Provincial Clinical Research Center for Child Health, Guangzhou, 510623, People’s Republic of China; 3The First Affiliated Hospital, Hengyang Medical School, University of South China, Hengyang, 421001, People’s Republic of China; 4Department of Rheumatology, Guangzhou First People’s Hospital, South China University of Technology, Guangzhou, 510623, People’s Republic of China; 5Department of Surgical Sciences, Anesthesiology and Intensive Care, Uppsala University, Uppsala, Sweden; 6Department of Medical Cell Biology, Unit for Integrative Physiology, Uppsala University, Uppsala, Sweden; 7Department of Physiology, School of Basic Medical Sciences, Zhejiang University School of Medicine, Hangzhou, 310058, People’s Republic of China

*These authors contributed equally to this work

Correspondence: Chun Tang; Xiaoping Liu, Email tangch28@mail.sysu.edu.cn; liu_xiaoping@gwcmc.org

**Purpose:** Methyltransferase like 1 (METTL1) regulates epitranscriptomes via the m7G modification in mammalian mRNA and microRNA. Systemic lupus erythematosus (SLE) is caused by abnormal immune reactivity and has diverse clinical manifestations. RNA methylation as a mechanism to regulate gene expression is widely implicated in immune regulation. However, the role of m7G in immune response of SLE has not been extensively studied.

**Patients and Methods:** Expression of METTL1 was identified in the public dataset GSE122459 and validated in an independent cohort of SLE patients. We investigated the association between METTL1-expression and clinical manifestations of SLE. Subsequently, differentially expressed genes (DEG) that were correlated with METTL1-expression in GSE122459 were used for functional enrichment analysis. The correlation between infiltrating immune cells and METTL1, as well as candidate biomarkers identified to be correlated with either METTL1 or immune cell infiltration were assessed by single-sample GSEA. Potential mechanisms were explored with Gene ontology and KEGG pathway enrichment. Diagnostic performances of candidate biomarkers in SLE were analyzed.

**Results:** The mRNA and protein expression of METTL1 in SLE patients were significantly decreased in both datasets. METTL1-coexpressed DEGs were enriched in several key immune-related pathways. Activated CD8 T cells, activated CD4 T cells, memory B cells and type 2 helper T cells were different between patients with high and low METTL1 expression. Further, activated CD8 T-cells, activated CD4 T-cells, memory B-cells were correlated with METTL1. The genes of LAMP3, CD83, PDCD1LG2, IGKVD3D-20, IGKV5-3, IGKV2D-30, IGLV3-19 and IGLV4-60 were identified as candidate targets that were correlated with immune cell proportion. Moreover, LAMP3, CD83, and PDCD1LG2 expression were of diagnostic value in SLE as indicated by ROC analysis.

**Conclusion:** Our findings suggested that METTL1 and its candidate targets LAMP3, CD83, PDCD1LG2 may be used for diagnosing SLE and could be explored for developing targeted molecular therapy for SLE.

**Keywords:** methyltransferase like 1, systemic lupus erythematosus, immunity, single sample gene set enrichment analysis, biomarker
Introduction
Systemic lupus erythematosus (SLE) is caused by production of pathogenic autoantibodies and has diverse clinical manifestations caused by the abnormal immune responses. Studies have shown phenotypic and functional abnormalities of immune cells, including T cells and B cells, in SLE patients. Multiple subsets of T cells of SLE patients show impaired cytokine production and deficient immune capacity. Besides, SLE patients and lupus-prone mice display various B cell abnormalities including B cell lymphopenia and B cell hyperactivity. Increasing studies have shown that innate immunity deficiency is identified as important factors in the pathogenesis of SLE. Although researches on lupus have made rapid advances, the explicit pathogenesis remains elusive. Conventional treatments are often accompanied with unwanted side-effects like infection and kidney damage, which seriously impair the quality of life of SLE patients. Hence, the mechanism behind clinical presentation and symptoms of SLE needs to be further elucidated in order to developing targeted and individualized therapy.

In the last decade, epigenetic mechanisms such as DNA methylation and histone modification have been identified during the development of SLE. Recently RNA methylation has been identified as a novel form of epigenetic regulation and reported to have effects in, for example, cancer and cardiovascular disease. Accumulating evidences show that RNA methylation is related to immune cell maturation and response, and its impacts on autoimmune disease is becoming apparent. RNA methylation is extensively distributed in multiple types of RNA, mainly comprising N7-methylguanosine (m7G), N6-methyladenosine (m6A), and 5-methylcytosine (m5C). Luo et al previously reported that multiple m6A-related enzymes including METTL14, ALKBH5 and YTHDF2 mRNA expression were decreased in PBMC of SLE patients, and this team also found that METTL3, METTL14, WTAP, FTO, ALKBH5 and YTHDF2 mRNA expression were decreased in PBMC of SLE patients, and ALKBH5 level was associated with anti-dsDNA, skin rash and ulcer, which can be a risk factor for SLE. Research also found that silencing of m5C demethyltransferase TET2 obviously inhibited Tfh cell polarization, which played a critical role in SLE. Methyltransferase like 1 (METTL1) is required to catalyze m7G, which is widely present in mRNA, tRNA, and rRNA. Recent studies have found that knockout of Mettl1 gene can induce defective self-renewal and neural differentiation of embryonic stem cells via disordered m7G tRNA modification. In addition, METTL1 has been shown to activate the let-7e miRNA/HMGA2 axis in a m7G-dependent manner. RNA methylation, like m6A and m5C, has been reported in SLE, but m7G has not been well characterized in SLE, and thus its function and mechanism need to be explored.

In this study, the METTL1 expression pattern of SLE was studied in a public dataset and an independent cohort of SLE patients. The relationship between METTL1 and immune cell infiltration in SLE was analyzed by single-sample Gene Set Enrichment Analysis (ssGSEA). Furthermore, a variety of promising METTL1-related immune molecules for the diagnosis of SLE were also investigated. Taken together, our study aims to understand the involvement of METTL1 in immune dysregulation during SLE.

Materials and Methods
Data Availability
The gene expression profiles were obtained from the Gene Expression Omnibus (GEO) public database (http://www.ncbi.nlm.nih.gov/geo/). GSE122459, consist of gene expression of peripheral blood mononuclear cells (PBMCs) from SLE patients and healthy controls (HC), and was used as the discovery dataset. In addition, the gene expression dataset GSE50772 was used for external validation.

SLE Patients and Clinical Information
Whole blood samples were obtained from 18 normal healthy controls and 45 eligible SLE patients, from the Seventh Affiliated Hospital of Sun Yat-sen University in Shenzhen, China. All cases met the SLE classification criteria of the American College of Rheumatology. Clinical information and laboratory examination data of SLE patients were collected and evaluated according to SLE Disease Activity Index (SLEDAI) scoring criteria. SLE patients with SLEDAI score ≥9 were considered to be active, while patients with SLEDAI score <9 were considered to be stable. This study was approved by the Ethics Committee of the Seventh Affiliated Hospital of Sun Yat-sen University and
conducted in accordance with the Declaration of Helsinki. All patients and healthy volunteers signed informed consent. PBMCs were isolated from freshly heparinized blood samples using human peripheral blood lymphocyte separation medium (P8610, Solarbio, Beijing, China) according to the manufacturer’s instructions.

**Identification of Immune-Related Differentially Expressed Genes (ImmuneDEGs)**

METTL1-coexpressed genes by Pearson correlation test in GSE122459 were selected. The threshold points for METTL1-coexpressed genes were absolute $r > 0.4$ and $P < 0.05$. ImmPort database (https://www.immport.org) or previous research was utilized to identify the immune-related genes. $^{32}$ DEGs overlapping with the METTL1-coexpressed genes and immune-related genes were defined as METTL1-coexpressed DEGs and immuneDEGs, respectively. Then, 8 immuneDEGs co-expressed with METTL1 were intersected, which were visualized by Venn diagram.

**Gene Set Enrichment Analysis (GSEA)**

In attempt to investigate the potential mechanism of METTL1 in SLE progression, GSEA between low and high METTL1 expression group was conducted. Molecular Signatures Database (MSigDB) KEGG gene set was selected as the reference file. The NES, NOM $P$-value and FDR were used to quantify the statistically significant enrichment.

**Gene Functional Enrichment Analysis**

The enrichment analysis of METTL1-coexpressed DEGs and ImmuneDEGs was conducted using Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) in the DAVID database (https://david.ncifcrf.gov/). GO enrichment analysis included biological process (BP), molecular function (MF), and cellular component (CC). An enrichment $P$ value $<0.05$ was considered significant.

**Immune Cells Infiltration Analysis**

The ssGSEA was applied to analyze the abundance of immune cells in these patients. According to immune cell-specific marker genes, the enrichment score of the gene set represented the proportion of 28 types of infiltrating immune cells. The ssGSEA analysis was performed using the R package. The patients with METTL1 low- and high-expression groups were defined in the above dataset based on median METTL1 expression values. Subsequently, differential infiltration levels of immune cells were compared between the two groups.

**Correlation Analysis Between METTL1 and Immune Cells**

We further explored the association between METTL1 or the identified biomarkers and abundance of 28 immune cells. The correlation analyses were performed using Pearson coefficient and visualized in R software.

**Evaluation of the Diagnostic Effectiveness of Biomarkers**

The diagnostic values of the identified biomarker in patients with SLE were assessed using receiver operating characteristic curve (ROC) analysis by SSPA Statistics 25. The diagnostic accuracy of SLE versus HC was determined by the area under the ROC curve (AUC).

**Western Blotting**

For protein extraction, PBMCs were lysed by RIPA buffer. Protein lysates were quantified by the BCA method (23227, ThermoFisher Scientific, Vienna, Austria) and boiled with loading buffer. To detect the expression of METTL1, proteins were separated using SDS-PAGE and electrotransferred to PVDF membrane. After 1 h of blocking with 5% non-fat milk, membranes were followingly incubated with primary antibodies for anti-METTL1 (1:1000, PA5-80810, Invitrogen, Vienna, Austria) and anti-Vinculin (1:1000, 66305-1-Ig, Proteintech, Wuhan, China) at 4°C overnight. Post rinsing with TBST, the membranes were incubated with anti-rabbit IgG-HRP antibody (1:5000, SA00001-1, Proteintech, Wuhan, China) or anti-mouse IgG-HRP antibody (1:5000, SA00001-2, Proteintech, Wuhan, China) for 1 h at room temperature. Quantification was performed by measuring the intensity of the gels using ImageJ.
RNA Isolation and Real Time PCR

Total cellular RNA was extracted from PBMCs samples using TRIzol (108–95-2, Invitrogen) method. Complementary DNA (cDNA, 1 μg) was reverse transcribed from RNA using the PrimeScript RT reagent kit (RR047B, Takara, Beijing, China), and amplified in triplicate with TB Green Premix Ex Taq II (RR820B, Takara, Beijing, China) with specific primers by quantitative real-time PCR reactions. The primers used were as follows: METTL1 (Forward primer) 5’ AAACCCGGACTGGGATCAT 3’; METTL1 (Reverse primer) 5’ CTGGCTTCACAGGGTAGCG 3’; GAPDH (Forward primer) 5’ AGAAGGCTGGGGCTCATTTG 3’; GAPDH (Reverse primer) 5’ AGGGGCCATCCACAGTCTTC 3’; The relative abundance of mRNA was standardized against the levels of GAPDH as the internal control.

Statistical Analysis

Continuous variable analysis was conducted using unpaired t-test, or Mann–Whitney test for non-parametric data; Clinical features between groups were compared by a chi-squared test of independence. Correlation was performed using Pearson correlation test. P-value <0.05 was considered statistically significant. The statistical analyses were performed using R version 4.1.2, GraphPad Prism 7.0 and SPSS version 25.0 software.

Results

METTL1 Was Downregulated in SLE PBMCs

We conducted the study according to the following flow diagram (Figure 1). Using GSE122459 dataset, we found that the METTL1 mRNA expression was statistically downregulated in SLE compared with healthy controls (Figure 2A). To explore whether METTL1 plays a key role in SLE, we examined METTL1 expression in an independent SLE cohort, including a total of 45 SLE patients and 18 HC for further validation. Similarly, we revealed the mRNA level of METTL1 in PBMCs of SLE patients was remarkably lower than that of HC in the validation cohort. We classified the
patients into SLE patients with active disease (n = 20; SLEDAI score: 18.60 ± 2.63) and SLE patients in stable period (n = 25; SLEDAI score: 3.44 ± 0.65) using the SLEDAI score. Notably, we found that the mRNA expression of METTL1, among patients with active disease but not in stable period, were remarkably decreased compared to that in HC group (Figure 2B). Our data showed accordant change of METTL1 protein level by Western blotting (Figure 2C and D). In short, these results demonstrated that SLE patients exhibited a distinct expression of METTL1 as compared to healthy individuals.

METTL1 Was Correlated with B Cells and Suppressor T Cells in SLE
To explore the relationship between the expression of METTL1 and clinical features in SLE patients, we classified the samples into high and low METTL1 expression groups using the median METTL1 mRNA value. The clinicopathological characteristics of our cohort were selected and analyzed (Additional file 1: Table S1). However, there were no differences between SLE patients with high and low METTL1-level group after using Bonferroni to correct the p-values (Table 1). Notably, METTL1 was positively correlated with peripheral blood B cells, while negatively correlated with suppressor T cells in SLE (Figure 3A-L). These data suggested that the deficiency of METTL1 had the potential to participate in the progression of SLE.

METTL1 Was Enriched in Aberrant Immune Pathways in SLE
To gain insight into the biological processes of METTL1 in SLE, the GSEA analysis was used to predict the underlying signaling pathways of METTL1. The results showed that METTL1 expression was negatively correlated with Notch-signaling pathway, Fcγ-R-mediated phagocytosis, MAPK pathway and toll-like receptor signaling pathway, arousing our interest in the molecular network about METTL1 in SLE (Figure 4A and B).
Furthermore, we identified 5515 METTL1-coexpressed genes in the discovery dataset, including 3937 positively and 1578 negatively correlated co-expressed genes. Out of these, we defined 448 up-regulated and 60 down-regulated genes in SLE as DEGs, and obtained 35 METTL1-coexpressed DEGs (Figure 4C, additional file 2: Table S2). To explore enriched GO and KEGG pathways among the METTL1-coexpressed DEGs, we used DAVID Bioinformatics Resources and revealed that various immune-related biological pathways were significantly associated with METTL1-coexpressed DEGs. Our analysis showed that the enriched pathways mainly included immunoglobulin production, immunoglobulin complex, adaptive immune response, immune response and Fanconi anemia pathway (Figure 4D and E). Overall, our findings indicated that the function of METTL1 was closely related to the aberrant immune pathways known to be important in SLE.

### METTL1 Was Related to Immune Cell Infiltrates

Inspired by the pathway enrichment analysis results, we analyzed the infiltration of 28 immune cell subpopulations by ssGSEA in GSE122459 dataset and determined the fraction of 28 kinds of immune cells in each sample (Additional file 3: Table S3). We found 4 types of immune cells were significantly different between low and high METTL1 expression groups, presented in the violin diagram (Figure 5A). The results indicated that central memory CD4 T cells, activated CD8 T cells and monocytes were the main infiltrating immune cells. Then we evaluated the correlation of 28 types of immune cells in SLE PBMCs. Data were shown that activated CD8 T cells were positively correlated with activated CD4 T cells.

#### Table 1 Comparison of Clinical Characteristics Between SLE Patients with METTL1 Low and High Expression

<table>
<thead>
<tr>
<th></th>
<th>Patients with METTL1 mRNA Low Expression (n=23)</th>
<th>Patients with METTL1 mRNA High Expression (n=22)</th>
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<tr>
<td>Female, n (%)</td>
<td>19 (82.61)</td>
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<td>Vasculitis, n (%)</td>
<td>4 (17.39)</td>
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<td>Myositis, n (%)</td>
<td>1 (4.35)</td>
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<td>Fever, n (%)</td>
<td>7 (30.43)</td>
<td>5 (22.73)</td>
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<td>Hypocomplementemia, n (%)</td>
<td>15 (65.22)</td>
<td>10 (45.45)</td>
<td>0.182</td>
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<tr>
<td>Pericarditis, n (%)</td>
<td>4 (17.39)</td>
<td>1 (4.55)</td>
<td>0.170</td>
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<td>Pleuritis, n (%)</td>
<td>2 (8.70)</td>
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<td>Oral ulcer, n (%)</td>
<td>3 (13.04)</td>
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<td>Pneumonia, n (%)</td>
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<td>1 (4.55)</td>
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<td>Pyuria, n (%)</td>
<td>3 (13.04)</td>
<td>5 (22.73)</td>
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<tr>
<td>Proteinuria, n (%)</td>
<td>6 (26.09)</td>
<td>10 (45.45)</td>
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<td>Hematuria, n (%)</td>
<td>6 (26.09)</td>
<td>3 (13.64)</td>
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<td>Cylindruia, n (%)</td>
<td>4 (17.39)</td>
<td>0 (0)</td>
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<td>Renal involvement, n (%)</td>
<td>10 (43.48)</td>
<td>9 (40.91)</td>
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<td>0 (0)</td>
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<td>Arthritis, n (%)</td>
<td>9 (39.13)</td>
<td>6 (27.27)</td>
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<td>Skin rash, n (%)</td>
<td>8 (34.78)</td>
<td>5 (22.73)</td>
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<td>Anti-dsDNA, n (%)</td>
<td>7 (30.43)</td>
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<td>Anti-Sm, n (%)</td>
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<td>SLEDAI</td>
<td>13.52±13.46</td>
<td>6.68±6.61</td>
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</table>
central memory CD4 T cells and memory B cells, while negatively related to monocytes (Figure 5B). These results revealed a distinct spectrum of immune cell levels among low and high METTL1 expression group in SLE patients.

Additionally, we further examined the correlations between METTL1 expression and immune cells using the GSE122459 dataset (Figure 5B, additional file 4: Table S4). Our analysis indicated that METTL1 was positively correlated with central memory CD4 T cell, activated CD8 T cell, immature B cell, activated B cell, activated CD4 T cell and memory B cells, while negatively correlated with T follicular helper cell, regulatory T cell, monocyte, gamma delta T cell, and macrophage (Figure 5B and C). These results suggested that METTL1 might affect the dysregulation of immune cells in SLE.

Figure 3 Relationship between downregulation of METTL1 and clinical features in SLE. Correlation between METTL1 expression and age (A), C3 (B), C4 (C), IgM (D), IgA (E), IgG (F) (n=45), peripheral T lymphocyte (G), T helper cell (H), suppressor T cell (I), B lymphocyte (J) (n=12), NK cell (K) (n=10), and CD4+CD8+ (L) (n=14). Pearson correlation test.

Abbreviations: METTL1, Methyltransferase like 1; SLE, Systemic lupus erythematosus.
Figure 4 Exploration of METTL1-related function enrichment analysis. (A) GSEA enriched pathways in low METTL1 expression group; (B) Multiple GSEA enriched pathways in low METTL1 expression group; Volcano plot (C) for 35 METTL1-coexpressed DEGs; Bubble plot (D) and Circos plot (E) for the enrichment of METTL1-coexpressed DEGs.

Abbreviations: METTL1, Methyltransferase like 1; SLE, Systemic lupus erythematosus; GSEA, Gene Set Enrichment Analysis; METTL1-coexpressed, DEGs METTL1-coexpressed differentially expressed genes.
Intersection and Validation of ImmuneDEGs Co-Expressed with METTL1

SLE is characterized by aberrant immune response and release of inflammatory mediators. To determine ImmuneDEGs in SLE, we compared DEGs with the immune-related genes and identified 66 ImmuneDEGs (Additional file S5: Table S5). By analyzing the GO functions of these ImmuneDEGs, we observed that they were mainly enriched in the immunoglobulin
production and immune response (Figure 6A and B). Also, KEGG analysis revealed that IL-17 signaling pathway, chemokine signaling pathway and cytokine–cytokine receptor interaction pathways were significantly enriched (Figure 6C and D).

To further explore the possible immune targets of METTL1 in SLE, we screened 8 ImmuneDEGs that simultaneously co-expressed with METTL1 via venn diagram, including lysosomal associated membrane protein 3 (LAMP3), CD83 molecule (CD83), programmed cell death 1 ligand 2 (PDCD1LG2), IGKVD3D-20, IGKV5-2, IGKV2D-30, IGLV3-19 and IGLV4-60 (Figure 6E). And we visualized the final candidate biomarkers by the volcano map (Figure 6F) and heatmap (Figure 6G). Afterwards, we verified the expression trend of LAMP3, CD83 and PDCD1LG2, which agreed with that in GSE50772 (Figure 6H). These results suggested that these candidate genes might be downstream targets of METTL1 and involved in abnormal immune response of SLE.

**METTL1-Coexpressed ImmunDEGs Correlated with Infiltrating Immune Cells**

Subsequently, we analyzed the correlation of LAMP3, CD83, PDCD1LG2, IGKVD3D-20, IGKV5-2, IGKV2D-30, IGLV3-19 and IGLV4-60 with various immune cell types in SLE using the GSE122459 dataset (Additional file 6: Table S6). We found that LAMP3 was significantly correlated with activated CD8 T cell and activated CD4 T cell (Figure 7A); CD83 was remarkably correlated with activated CD8 T cell (Figure 7B); PDCD1LG2 was not correlated with activated CD8 T cell, activated CD4 T cell or memory B cell (Figure 7C). IGLV3-19, IGKV5-2 and IGKV4-60 was greatly positively correlated with activated CD8 T cell, activated CD4 T cell and memory B cell, respectively (Figure 7D-F); IGKVD3D-20 was evidently positively correlated with activated CD4 T cell and memory B cell (Figure 7G); IGKV2D-30 was remarkably positively correlated with activated CD4 T cell (Figure 7H). To our interest, these genes were mostly related to the activated CD8 T cell, activated CD4 T cell and memory B cells, which were distinctly expressed among low and high METTL1 groups. In brief, METTL1 and these co-expressed ImmunDEGs may collectively participate in the immune pathogenesis of SLE by regulating the immune cells.

**METTL1 and Coexpressed ImmunDEGs Had Favorable Diagnostic Values in Both Discovery and Validated Cohorts**

To evaluate the diagnostic performance of METTL1, LAMP3, CD83, PDCD1LG2, IGKVD3D-20, IGKV5-2, IGKV2D-30, IGLV3-19 and IGLV4-60 in SLE, we performed ROC analysis using GSE122459 dataset. The results showed that the AUC values of METTL1, LAMP3, CD83 and PDCD1LG2 in SLE were respectively 0.800, 0.900, 0.883 and 0.825, with the combined AUC of LAMP3, CD83 and PDCD1LG2 reaching 0.975 (Figure 8A and B). The AUC values of IGKVD3D-20, IGKV5-2, IGKV2D-30, IGLV3-19 and IGLV4-60 in SLE were respectively 0.850, 0.783, 0.858, 0.858 and 0.740, with the combined AUC of IGKVD3D-20, IGKV5-2, IGKV2D-30, IGLV3-19 and IGLV4-60 reaching 0.925 (Figure 8C). The above data indicated that these candidate biomarkers had high accuracy in classifying SLE patients. Afterwards, we employed our SLE cohort and GSE50772 dataset to validate the results. The AUC values of METTL1 in our SLE cohort were 0.655 (Figure 8D). We also revealed that the diagnostic ability of the three genes was consistent with the results from the GSE122459 dataset, among which the diagnostic power of LAMP3 was better than that of CD83, and PDCD1LG2 (AUC of LAMP3, CD83, PDCD1LG2 and the three combined: 0.795, 0.670, 0.657 and 0.832) (Figure 8E). These results indicate that the candidate genes have favorable diagnostic performance for SLE.

**Discussion**

The main result of the present study was that METTL1 was down regulated in SLE and that putative target genes are associated with immune cell activation consistent with the SLE phenotype. Epigenetic regulation, such as DNA methylation and histone modification, has been found to affect the development and function of the immune system by regulating gene expression, thus participating in the occurrence of autoimmune diseases. Increasing recent evidence suggests that RNA methylation plays an important role in the pathogenesis of SLE. Several studies have shown that m6A modification levels were dysregulated in SLE patients and closely related to the regulation of lncRNA expression. In addition, multiple m6A-associated enzymes, including METTL3, METTL14, WTAP, FTO, ALKBH5 and YTHDF2,
Figure 6 Intersection and validation of ImmuneDEGs co-expressed with METTL1. Bubble plot (A) and circos plot (B) for the GO enrichment analysis of ImmuneDEGs; Bubble plot (C) and circos plot (D) for the KEGG enrichment of ImmuneDEGs; Venn diagram (E) of 8 candidate genes overlapping ImmuneDEGs and METTL1-coexpressed genes; Volcano plot (F) and Heatmap (G) for 8 candidate genes between SLE and HC; Violin plot (H) for the expression of LAMP3, CD83 and PDCD1LG2 in GSE50772 dataset for validation. Mann–Whitney test. Data are means ± SEM. *P < 0.05, ***P < 0.001.

Abbreviations: ImmuneDEGs, Immune-related differentially expressed genes; METTL1, Methyltransferase like 1; SLE, Systemic lupus erythematosus; GO, Gene Ontology; BP Biological process; MF, Molecular function; CC, Cellular component; KEGG, Kyoto Encyclopedia of Genes and Genomes; M-co, DEGs METTL1-coexpressed differentially expressed genes; HC, Healthy controls; LAMP3, Lysosomal associated membrane protein 3; CD83, CD83 molecule; PDCD1LG2, Programmed cell death 1 ligand 2.
Figure 7 Correlation between METTL1-coexpressed ImmuneDEGs and differential expressed immune cells. Correlation between LAMP3 (A), CD83 (B), PDCD1LG2 (C), IGKV5-2 (D), IGKV3D-20 (E), IGLV4-60 (F), IGLV3-19 (G), IGKV2D-30 (H) and infiltrating immune cells. Pearson correlation test.

Abbreviations: METTL1, Methyltransferase like 1; ImmuneDEGs, Immune-related differentially expressed genes; LAMP3, Lysosomal associated membrane protein 3; CD83, CD83 molecule; PDCD1LG2, Programmed cell death 1 ligand 2.
as well as m5C methyltransferase NSUN2, were dysregulated in SLE patients. Recent studies revealed that METTL1 may regulate epitranscriptomes via the m7G modification in the 3′UTR of mammalian mRNA and microRNA. METTL1-mediated m7G is known for diverse processes of RNA metabolism, such as the regulation of mRNA transcription, microRNA (miRNA) biosynthesis, and tRNA stability. Although METTL1 has been reported to drive the progression of various diseases, there are no studies addressing its role in autoimmune disorders, such as SLE. Our study found that METTL1 expression was both transcriptionally and translationally downregulated in SLE patients. Previous studies have found that the expression of m6A- and m5C-related proteins in SLE patients was dysregulated. Our study further confirmed that m7G methyltransferase METTL1 was also dysregulated in SLE patients, which indicated m7G may be involved in the development of SLE.

In addition, recent studies have found that liver-specific overexpression or knockdown of Mettl1 distinctly influenced the accumulation of polymorphonuclear-myeloid-derived suppressor cells, which consequently affected the infiltration of CD8+T cells. METTL1 also exhibited a significant association with tumor immune microenvironment infiltration and various immune modulators. However, studies on the role of METTL1 in the immune system and autoimmune diseases are still limited. Our study found that abnormal expression of METTL1 in SLE was closely related to the abundance of multiple immune cells. METTL1 level was related to peripheral blood B cells and suppressor T cells. To explore its potential function in SLE, we employed GSEA analysis. The data demonstrated that genes were more likely to be enriched in Notch signaling pathway, Fcγ-R-mediated phagocytosis, MAPK pathway and toll-like receptor signaling pathway in SLE cohorts with low METTL1 expression. All these pathways were reported to be involved in SLE by

Figure 8 Diagnostic values of METTL1-coexpressed ImmuneDEGs in both discovery and validated cohorts. (A) ROC analysis for METTL1 in GSE122459; (B) ROC analysis for LAMP3, CD83, and PDCD1LG2 in GSE122459; (C) ROC analysis for IGKV5-2, IGKV3D-20, IGLV4-60, IGLV3-19, and IGKV2D-30 in GSE122459; (D) ROC analysis for METTL1 in our SLE cohort; (E) ROC analysis for LAMP3, CD83, and PDCD1LG2 in GSE50772.

Abbreviations: METTL1, Methyltransferase like 1; ImmuneDEGs, Immune-related differentially expressed genes; LAMP3, Lysosomal associated membrane protein 3; CD83, CD83 molecule; PDCD1LG2, Programmed cell death 1 ligand 2; ROC, Receiver operating characteristic curve.
regulating immune cells or inflammatory cytokines. The GO and KEGG analysis indicated that METTL1-coexpressed DEGs were enriched in immune-related pathways enrichment. To sum up, these results suggested that METTL1 may play a key role in SLE immunity.

The aberrant subpopulations and function of T and B lymphocytes have been identified as critical factors in the pathophysiology of SLE. The effects of RNA methylation on the immune system have been extensively explored, including differentiation and maturation of T cells, early B cell development, and macrophage activation. By analyzing ssGSEA data, we revealed an immune infiltration in SLE, and found that the fraction of activated CD8 T cells, activated CD4 T cells, Type 2 T helper cells and memory B immune cells were remarkably varied among groups with low and high expression of METTL1. Besides, we explored the relationship between METTL1 and immune cells infiltration for the possible immune-related value of METTL1. The results revealed the marked positive correlativity between METTL1 and activated CD8 T cell, activated CD4 T cell, memory B cells. There has been reported that CD4+ and CD8+ T cells in SLE were abnormally activated on account of aberrant T cell signaling pathways, thus leading to inflammatory responses, help to B cells, and autoimmunity. It has been found that increased CD28-deficient CD8+ T cells and switched memory B cells were correlated with disease activity of SLE patients. Therefore, we supposed that METTL1 could influence the immunity of SLE by regulating these aberrant immune cells.

Considering the essential role of abnormal immune system in the pathophysiology of SLE, the dysregulated immune-related genes are promising targets for modulating its progression. We identified eight DEGs: LAMP3, CD83, PDCD1LG2, IGKV3D-20, IGKV5-2, IGLV3-19 and IGLV4-60 that overlapped with both the METTL1-coexpressed DEGs and ImmuneDEGs. Our analysis showed that LAMP3 was significantly positively correlated with activated CD8 T cell and activated CD4 T cell; CD83 was remarkably positively correlated with activated CD8 T cell; IGLV3-19, IGKV5-2 and IGKV4-60 was greatly positively correlated with activated CD8 T cell, activated CD4 T cell and memory B cell, respectively; IGKV2D-30 was remarkably positively correlated with activated CD4 T cell; IGKVD3D-20 was greatly positively correlated with activated CD4 T cell and memory B cell. Previous research reported that LAMP3+ dendritic cells can affect depleted CD8+ T cells through PDL1-PD1 and CD200-CD200R signaling. The abundance of LAMP3+ dendritic cells in tumors may be positively correlated with the infiltration levels of functionally depleted CD8+ T cells and CD4+ regulatory T cells. In addition, the downregulation of CD83 on dendritic cells was revealed to decrease capacity in the priming of functional tumor antigen-specific CD8+ T cells. Moreover, variable (IGKV and IGLV) genes encoded immunoglobulin kappa (IGK) and immunoglobulin lambda (IGL) light chain, which were both B cell receptor components closely related to BCR signaling pathways. From all above, these data indicated METTL1 and its candidate genes could be conjointly involved in the activation of immune response by affecting various immune cells.

For ROC curve analysis, we found that LAMP3, CD83, and PDCD1LG2 were of diagnostic values in SLE, and the diagnostic value of LAMP3 was much higher than CD83 and PDCD1LG2. While the diagnostic performances of LAMP3, CD83 and PDCD1LG2 have been described in various cancers, their diagnostic values in SLE have not been reported previously. Our findings provided additional candidate diagnostic biomarkers for SLE. However, considering the limitation of clinical samples of patients, the expression levels, and diagnostic performance of LAMP3, CD83, and PDCD1LG2 in SLE require independent validation.

**Conclusion**

In summary, the present study suggested that reduced expression of m7G methyltransferase METTL1 was related to peripheral blood B cells and suppressor T cells in SLE. We identified that METTL1 and its co-expressed ImmuneDEGs LAMP3, CD83, and PDCD1LG2 had strong correlations with immune cell abundance. Moreover, we demonstrated that LAMP3, CD83, and PDCD1LG2 were promising diagnostic biomarkers for SLE. The data in this study provide new insights for key proteins in RNA methylation and novel diagnostic biomarkers in SLE with potential mechanistic roles in SLE development that can inform future research.

**Abbreviations**

SLE, Systemic lupus erythematosus; m7G, N7-methylguanosine; m6A, N6-methyladenosine; m5C, 5-methylcytosine; METTL1, Methyltransferase like 1; WDR4, WD repeat domain 4; ssGSEA, Single-sample gene set enrichment analysis;
GEO, Gene Expression Omnibus; PBMCs, Peripheral blood mononuclear cells; HC, Healthy controls; DEGs, Differentially expressed genes; SLEDAI, SLE Disease Activity Index; ImmuneDEGs, Immune-related differentially expressed genes; GSEA, Gene Set Enrichment Analysis; GO, Gene Ontology; BP, Biological process; MF, Molecular function; CC, Cellular component; KEGG, Kyoto Encyclopedia of Genes and Genomes; ROC, Receiver-operating characteristic curve; AUC, Area under the ROC curve; LAMP3, Lysosomal associated membrane protein 3; CD83, CD83 molecule; PDCD1LG2, Programmed cell death 1 ligand 2.

Data Sharing Statement
The datasets supporting the conclusions of this study are available in the Gene Expression Omnibus (GEO) datasets (http://www.ncbi.nlm.nih.gov/geo/) and its additional files.

Study Approval
This study was approved by the Research and Experimental Animal Ethics Committee of the Seventh Affiliated Hospital of Sun Yat-sen University (Certificate Number: 2019SYSUSH-037) and conducted in accordance with the Declaration of Helsinki.

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Author Contributions
All authors made a significant contribution to the work reported, whether that is in the conception, study design, execution, acquisition of data, analysis and interpretation, or in all these areas; took part in drafting, revising or critically reviewing the article; gave final approval of the version to be published; have agreed on the journal to which the article has been submitted; and agree to be accountable for all aspects of the work.

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Disclosure
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