

Disulfidptosis-related gene SLC7A11 predicts prognosis and indicates tumor immune infiltration in lung adenocarcinoma

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> **Background:** Lung adenocarcinoma (LUAD) is closely associated with factors such as smoking and metabolic disorders. A unique form of cell death known as disulfidptosis, which is regulated by genes like *SLC7A11*, has emerged as an area of interest; however, its effect on the immune microenvironment in the context of cancer remains largely unexplored. The aim of this study was to analyze the immunoregulatory role of disulfidptosis-related genes in LUAD to unveil and underscore their significance in the process of immune regulation.

> Methods: This study examined the role of disulfidptosis-related genes in LUAD using data from The Cancer Genome Atlas (TCGA) with a particular focus on immune infiltration and the function of *SLC7A11*. The research employed a clustering analysis, survival analysis, and immune function assessment, integrating both bulk and single-cell RNA sequencing data, to gain a comprehensive understanding of disulfidptosis in LUAD.

> Results: The analysis revealed three distinct LUAD clusters, each characterized by different survival rates and patterns of immune cell infiltration. Notably, high expression levels of *SLC7A11* were associated with a poor prognosis and mechanisms of immune evasion. High SLC7A11 expression is correlated with a poor prognosis and immune evasion in LUAD. These results underscore the significant role of *SLC7A11* in the progression of disulfidptosis and LUAD.

> **Conclusions:** This study sheds new light on the role of disulfidptosis in LUAD, particularly highlighting the immunoregulatory effects of *SLC7A11*. The findings suggest that targeting *SLC7A11* could lead to the development of novel therapeutic strategies aimed at enhancing the response to immunotherapy in LUAD patients. To substantiate these results, further experimental validation is needed.

Keywords: Lung adenocarcinoma (LUAD); disulfidptosis; tumor microenvironment (TME); immunotherapy

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Introduction

Lung cancer is the most common cancer and leading cause of cancer-related death worldwide (1,2). Non-small cell lung cancer (NSCLC) represents approximately 85% of all lung cancer cases, and can be further classified into lung adenocarcinoma (LUAD), squamous cell carcinoma, and large cell carcinoma (3). Of these, LUAD is the most frequently occurring NSCLC subtype, and its

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development is linked to smoking, alcohol consumption, and metabolic imbalances (4,5). Despite advancements in LUAD treatments, which include surgery, chemotherapy, radiotherapy, targeted therapy, and immunotherapy, patient survival rates remain disappointingly low (6). Recent research has highlighted the insufficiency of traditional histological categorizations of LUAD, due to the inherent heterogeneity and complex nature of the disease (7). This has led to an increased focus on molecular subtype research to refine treatment approaches. Understanding how genetic alterations disrupt the tumor microenvironment (TME) and affect LUAD prognosis is crucial for identifying novel therapeutic targets (8,9).

Disulfidptosis is a novel form of cell death distinct from other known types such as apoptosis and ferroptosis (10,11). It occurs under certain conditions, such as glucose starvation in cells with high expression of the *SLC7A11* gene (12,13). The other genes were *FLNA*, *FLNB*, *MYH9*, *TLN1*, *ACTB*, *MYL6*, *MYH10*, *CAPZB*, *DSTN*, *IQGAP1*, *ACTN4*, *PDLIM1*, *CD2AP*, and *INF2*, which are closely related to disulfidptosis. This process is characterized by the accumulation of intracellular disulfides, leading to aberrant disulfide bonds in cytoskeleton proteins, which in turn affect the actin cytoskeleton, causing cell death. That

Highlight box

Key findings

- Three distinct clusters in lung adenocarcinoma (LUAD) were identified based on disulfidptosis-related genes.
- Elevated expression of *SLC7A11* is associated with poor prognosis and immune evasion in LUAD.
- Disulfidptosis likely plays a significant role in LUAD progression and immune regulation.

What is known, and what is new?

- LUAD is heterogeneous and is influenced by factors such as smoking. Disulfidptosis, a novel form of cell death, is influenced by genes like *SLC7A11*, but little is known about its role in cancer immunity.
- This study identified LUAD clusters based on disulfidptosis genes and linked high *SLC7A11* expression to a poor prognosis and immune evasion, offering insights into LUAD immunobiology.

What is the implication, and what should change now?

• An understanding of disulfidptosis-related genes in LUAD may lead to the development of novel therapies. Our findings suggest that *SLC7A11* could serve as a potential target to enhance immunotherapy. Further research should focus on developing *SLC7A11*-targeted therapies and incorporating *SLC7A11* status in patient stratification for personalized immunotherapy.

is caused by disulfide stress caused by the accumulation of excess cystine in cells. The study of genes related to disulfide death is of great significance for understanding the mechanism of cell death and exploring potential cancer therapies. Current research on the role of disulfidptosis in cancer is burgeoning, yet there are no reports on its effect on the immune microenvironment (13). Our study aimed to analyze the immunoregulatory role of disulfidptosis-related genes in LUAD to unveil and underscore their significance in the process of immune regulation.

In this study, we examined 15 disulfidptosis-related genes previously reported (14) and conducted a clustering analysis of the sequencing results of the LUAD cohort from The Cancer Genome Atlas (TCGA). Subsequently, we analyzed the immune infiltration scores of the clusters regulated by disulfidptosis. By integrating the results of bulk and singlecell RNA sequencing from immune therapy cohorts, our research findings shed light on the immunoregulatory role of disulfidptosis in cancers, exemplified by LUAD. We present this article in accordance with the STREGA reporting checklist (available at [https://tcr.amegroups.com/](https://tcr.amegroups.com/article/view/10.21037/tcr-24-1182/rc) [article/view/10.21037/tcr-24-1182/rc\)](https://tcr.amegroups.com/article/view/10.21037/tcr-24-1182/rc).

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Preprocessing the RNA expression data

The RNA expression datasets for TCGA-LUAD were meticulously compiled using the UCSC Xena platform (https://xenabrowser.net/datapages/). A total of 511 cases were included in the study. Additionally, comprehensive somatic mutation data corresponding to these cases were acquired from the Genomic Data Commons (GDC) portal of the National Cancer Institute (NCI), which can be accessed at the NCI GDC Repository ([https://portal.gdc.cancer.](https://portal.gdc.cancer.gov/) [gov/\)](https://portal.gdc.cancer.gov/). Single-cell sequencing data from patients treated with programmed cell death protein 1 (PD1) therapy were curated from a breast cancer study (15). Such datasets are invaluable in advancing understandings of the molecular intricacies in LUAD. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013).

Principal component analysis (PCA) and consensus molecular clustering by non-negative matrix factorization (NMF)

A PCA, a critical technique in data dimensionality reduction, was conducted using highly variable genes. NMF, a method extensively employed for clustering in

high-dimensional datasets, especially in the realm of computational biology, was then applied. Focusing on TCGA-LUAD cohort, we identified distinct disulfidptosisrelated molecular clusters using a consensus clustering approach via the "NMF" function.

Data analysis for Clustered Regularly Interspaced Short Palindromic Repeats (CRISPR) screens

The CRISPR screen data were curated from lung cancer mouse *in vitro* and *in vivo* models (16). To process and analyze the CRISPR screen data, we employed MaGeCK software. Specifically, Fastq reads from the CRISPR screen underwent processing via the MaGeCK count module. Subsequently, using the default parameters, the robust rank aggregation (RRA) module was used to calculate the log₂fold changes and P values associated with the genes. Additionally, custom scripts in R (version 4.1.3) were developed and used to visualize the data.

Immune analysis

We applied gene signatures of immune cells, as identified by Charoentong *et al.* (17), to compute the scores related to immune infiltration. This computation was performed using the single-sample gene set enrichment analysis (ssGSEA) technique.

Statistical analyses

All the statistical analyses were conducted using R software (version 4.1.2) and GraphPad Prism (version 9.50). The Wilcoxon test, log-rank test, and Kruskal-Wallis *H* test were employed to evaluate the data. For comprehensive details regarding the statistical tests used, please refer to the specific annotations provided in the figure legends.

Results

Genetic variation of disulfidptosis-related genes in LUAD

To study the regulatory role of disulfidptosis in lung cancer, we referenced a gene signature of disulfidptosis reported in previous article (18). Using TCGA-LUAD RNA expression cohort, we conducted a PCA of cancer and adjacent nontumor tissues (*Figure 1A*). The results demonstrated that the genes related to disulfidptosis could be used to distinguish between the tumor and adjacent non-tumor tissues in LUAD. *Figure 1B* shows the frequency of gene mutations in this category; FLNA had the highest mutation frequency. The results of the copy number variation (CNV) analysis are presented in *Figure 1C*. Combining the expression levels of these genes in the tumor and adjacent non-tumor tissues in LUAD (*Figure 1D*), the copy number amplification of *ACTB* and *CD2AP* were consistent with their high expression levels in tumors, while the decreased copy number amplifications of *MYH10*, *DSTN*, and *INF2* were consistent with their low expression levels in tumors. This suggests that these genes play a significant regulatory role in LUAD; thus, the regulatory mechanisms of disulfidptosis in the development of cancer urgently need to be explored.

Disulfidptosis-related clusters in LUAD

To better explore the regulatory role of the disulfidptosis in LUAD, we adopted the classical categorization algorithm (i.e., NMF) (19,20) to subgroup the disulfidptosis-related gene clusters based on gene expression levels in LUAD. The NMF rank algorithm (*Figure 2A*) suggested that dividing the LUAD into three groups was appropriate. *Figure 2B* shows the heatmap of the consensus matrix. Therefore, we concluded that under the regulation of disulfidptosis, LUAD can be divided into three significant groups; that is, clusters 1–3. A subsequent survival analysis (*Figure 2C*) showed that cluster 1 had the worst prognosis, while cluster 3 had the best prognosis. The pathway ssGSEA (21,22) revealed that the cell cycle activity in cluster 1 was more robust, which to some extent confirmed the reason for the poorer prognosis of cluster 1 (*Figure 2D*). Conversely, cluster 3 expressed higher levels of immune activation pathways, which suggests that patients in this LUAD subgroup might be more suitable for immunotherapy. To further corroborate this hypothesis, we used a common set of immune-related genes for a GSEA. The results showed that the activated cluster of differentiation 8 (CD8) T cell GSEA score (i.e., the CD8 activation-related score) was highly expressed in both clusters 1 and 3 (*Figure 2E*). This increase in cluster 3 confirmed that it was an immuneactivated LUAD subgroup. However, the activation of CD8 T cells in the cluster1 subgroup, which had the poorest prognosis, might indicate the presence of factors leading to exhausted CD8 T cells, rendering it unresponsive to immunotherapy.

Figure 1 Genetic variation of disulfidptosis-related genes in LUAD. (A) PCA comparing normal and tumor samples. (B) Mutations in 15 disulfidptosis-related genes in LUAD. (C) Frequency of CNVs in 15 disulfidptosis-related genes in TCGA-LUAD samples. (D) Boxplot illustrating the expression levels of 15 disulfidptosis-related genes in normal versus tumor tissues. Significance levels are indicated as *, P<0.05; ***, P<0.001; NS, not significance; PCA, principal component analysis; N, normal sample; T, tumor sample; TMB, tumor mutational burden; CNV, copy number variation; LUAD, lung adenocarcinoma; TCGA, The Cancer Genome Atlas.

Analyses of TME infiltration of disulfidptosis-related clusters

Given that clusters 1–3 demonstrated potential mechanisms of immune infiltration linked to prognosis, we proceeded to investigate the expression levels of disulfidptosis-related genes in these clusters. Notably, we found that *SLC7A11* was significantly overexpressed in cluster 1 (*Figure 3A*). *SLC7A11*, a gene known for its strong association with disulfidptosis, has also been reported to be highly expressed in cancer (23). This finding aligned with our analysis, which suggested a poorer prognosis for cancer patients in cluster 1. High *SLC7A11* expression was correlated with an immunosuppressed state in patients; however, it did not necessarily reflect the inherent immune-regulatory functions of the gene. To explore this further, we analyzed CRISPR screen single-guide RNA sequencing data (24) from cocultures of OT1 mouse T cells and tumor cells (*Figure 3B*). Our analysis identified *SLC7A11* as a significant contributor to immune evasion. Additionally, we used PD1 bulk tissue sequencing results for our analysis. The initial findings showed that patients with high *SLC7A11* expression had a poorer prognosis than those with low *SLC7A11* expression (*Figure 3C*). Further, we observed that patients with higher *SLC7A11* expression demonstrated a greater tolerance to immunotherapy than those with lower *SLC7A11* expression (*Figure 3C*). This indicates that *SLC7A11*, which is a critical gene regulating disulfidptosis, plays a key role in immune evasion.

Immune function analysis of disulfidptosis-related clusters

The bulk tissue sequencing levels indicated high *SLC7A11*

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Figure 2 Disulfidptosis-related clusters in LUAD. (A) Display of the NMF rank survey. The optimal number of clusters identified was three (rank =3). (B) Consensus heatmap generated from NMF analysis. (C) Kaplan-Meier survival plot categorized by NMF clusters. P values were calculated using the log-rank test. (D) Analysis of the pathway activities associated with the three NMF clusters. (E) ssGSEA of immune cells across the three clusters. The statistical significance among the clusters was evaluated using the Kruskal-Wallis *H* test. Significance levels are denoted as *, P<0.05; **, P<0.01; ***, P<0.001; NS, not significance; NMF, non-negative matrix factorization; GSVA, gene set variation analysis; BARD1, BRCA1 associated RING domain 1; IFNA, interferon alpha; IFNG, interferon gamma; MHC, major histocompatibility complex; CTL, cytotoxic T lymphocyte; ssGSEA, single-sample gene set enrichment analysis; CD4, cluster of differentiation 4; CD8 cluster of differentiation 8; MDSC, myeloid-derived suppressor cells; LUAD, lung adenocarcinoma.

Figure 3 Analyses of TME infiltration of disulfidptosis-related clusters. (A) Boxplot depicting the expression patterns of 15 disulfidptosisrelated genes across three clusters. (B) CRISPR screen data analysis indicating that *SLC7A11* is a key gene in immune evasion. (C) Kaplan-Meier survival curve analysis coupled with an examination of PD1 response proportions. Significance levels are denoted as **, P<0.01; ***, P<0.001. FDR, false discovery rate; NR, no response; R, response; RCC, renal cell carcinoma; TME, tumor microenvironment; CRISPR, Clustered Regularly Interspaced Short Palindromic Repeats; PD1, programmed cell death protein 1.

expression in the immunotherapy non-responsive group; however, we could not exclude the influence of the expression levels of immune microenvironment cells. Thus, we subsequently introduced single-cell sequencing data from PD1-treated patients (15). The data we referenced used T cell receptor (TCR) clonal expansion as an indicator of patient response to PD1 treatment. Thus, clonal expansion ("E") indicated a response to immunotherapy, and no expansion ("NE") indicated a tolerance to immunotherapy. The analysis revealed that *SLC7A11* was highly expressed in the tumor cells in the NE group (*Figure 4A*). Additionally, we used the average expression level of epithelial cells to represent the *SLC7A11* expression in each sample, and the results were consistent with the single-cell analysis; that is, the average expression of *SLC7A11* in the tumor cells of the NE group was higher than that of the E group (*Figure 4B*). Following this, we categorized patients into high- and low-expression groups based on the median of the average expression level of *SLC7A11* in tumor cells. We found that patients with low SLC7A11 expression had a lower proportion of CD8⁺ T cell infiltration. In summary, *SLC7A11* plays a role in immune evasion (*Figure 4C*).

Discussion

In this study, we explored the transcriptomic landscapes shaped by disulfidptosis-related genes. We found that the LUAD cohort from TCGA could be stratified into three distinct clusters, each characterized by unique clinicopathological features. Notably, cluster 1 was characterized by the poorest survival outcomes and a high degree of immune cell infiltration. While cluster 3 was characterized by robust immune cell infiltration, a favorable prognosis, and the notable presence of immuneactivating cells like activated CD8 T cells. This study

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Figure 4 Immune function analysis of disulfidptosis-related clusters. (A) Feature-plot and (B) violin-plot depicting the differential expression of *SLC7A11* in the "E" and "NE" groups. (C) Boxplot illustrating the average expression levels of *SLC7A11* in the "E" and "NE" groups. (A) All the dots represent cells with SLC7A11 expression and darker colors indicates higher expression levels. Additionally, a boxplot showing the variance in CD8⁺ T cell infiltration between the high and low *SLC7A11* expression groups. Statistical significance across the groups was assessed using the Wilcoxon test. tSNE, t-distributed stochastic neighbor embedding; E, expansion; NE, no expansion; CD8, cluster of differentiation 8.

effectively mapped the intratumoral transcriptomic diversity as influenced by disulfidptosis, shedding light on the tumor TME infiltration patterns. Disulfidptosis-driven molecular subtyping can offer fresh perspectives in LUAD pathogenesis.

By exploring the expression of disulfidptosis-related genes in our three distinct clusters, we observed that *SLC7A11* was highly expressed in cluster 1. This study showed that the role of *SLC7A11* in cancer is integral to the process of disulfidptosis, which is a unique form of cell death. When expressed at high levels, particularly in the context of glucose starvation, *SLC7A11* leads to an excessive accumulation of intracellular disulfides (13). This accumulation causes aberrant disulfide bonding in actin cytoskeleton proteins, leading to cell death. A previous well-known study showed that disulfidptosis occurs in the *SLC7A11*-high condition of cancer (25). Further, it suggested that targeting disulfide stress, particularly in *SLC7A11*-high tumors, could be a promising therapeutic strategy in cancer treatment. In our research, we confirmed that *SLC7A11* was highly expressed in LUAD patients with a poor prognosis, reinforcing its significant role in promoting the progression of LUAD. However, our study differed notably from previous research in that it focused more on the role of high *SLC7A11* expression in regulating the immune microenvironment in LUAD tumors. We discovered that *SLC7A11* is a gene associated with immune evasion and shows resistance to immunotherapy. Therefore, we believe that targeting and inhibiting *SLC7A11* in LUAD patients could enhance the response to immunotherapy.

However, our study mainly focused on the bioinformatic analyses, which falls short of robust confirmation. Therefore, a more in-depth experimental validation of our findings is required.

Conclusions

In conclusion, the study provided new insights into the role of disulfidptosis in LUAD, particularly the immunoregulatory effects of *SLC7A11*. Our findings could lead to the development of novel therapeutic strategies targeting *SLC7A11* to enhance immunotherapy responses

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in LUAD patients.

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Footnote

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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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