

Original Article



Identification Of Disulfidptosis-Related Genes In Glioblastoma Combining Bioinformatics Techniques And Machine Learning

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Abstract:

Glioblastoma (GBM) is the most malignant brain tumor, and with high heterogeneity and subsequent temozolomide (TMZ) resistance, the traditional treatment are often suboptimal. Disulfidptosis, a newly proposed form of cell death, plays a crucial role in the progression of various tumors, and its involvement in glioblastoma (GBM) remains further exploration. In this study, we combined bioinformatics with multiple machine learning algorithms to identify diagnostic and prognostic biomarkers associated with disulfidptosis in glioblastoma (GBM). First, we collected datas from TCGA and GTEx databases, and performed differential analysis, univariate Cox regression, and survival analysis. Additionally, we applied four machine learning algorithms to identify six candidate genes. We further investigated the potential biological functions and signaling pathways associated with these genes through GSEA and GSVA analysis. Subsequently, we constructed a ceRNA network and a drug-target network to explore their potential complex regulatory mechanisms. Finally, the expression levels of these genes were validated using clinical samples. Results show that these six candidate genes identified memorably, enriching in the JAK-STAT3 signaling pathway, were upregulated in glioblastoma (GBM) and were closely associated with patient prognosis. These findings may contribute to developing the molecular mechanisms underlying glioblastoma (GBM) and offer valuable implications for clinical diagnosis and treatment.

Keywords: Bioinformatics, Glioblastoma, GBM, Disulfidptosis, JAK-STAT3, Diagnosis

1. Introduction

Gliomas originate from glial cells in the brain and are among the most common primary malignant tumors in the brain(Qin et al., 2025). They are well-known for their high heterogeneity and high mortality rate(Zhao et al., 2023). According to the World Health Organization, gliomas are classified into grades I, II, III, and IV based on their malignancy, with higher grades indicating a greater degree of malignancy. GBM, classified as grade IV, represents the most malignant form of glioma. Additionally, the unique metabolic processes of GBM, such as glycolytic metabolism (Warburg effect) (Zhang et al., 2025)and choline

metabolism(Ma et al., 2023), increase its danger. As a result, there is currently no effective treatment for GBM in clinical practice(Mandal, Banerjee, & Mandal, 2025). The first-line treatment involves surgical resection combined with adjuvant chemotherapy, but patient prognosis remains poor(Rodríguez-Camacho et al., 2022; Xu et al., 2025), with a median overall survival of only 4-18 months post-surgery(Tamai et al., 2022). Moreover, due to the development of drug resistance, the efficacy of temozolomide, an alkylating agent used as first-line treatment, has been increasingly limited(Lee et al., 2024).

Therefore, addressing these therapeutic challenges and identifying new effective treatments is crucial in clinical practice.

Disulfidptosis, a novel form of cell death distinct from ferroptosis, apoptosis, and other types, has been proposed in recent years. It is caused by excessive intracellular disulfide stress, leading to cell death (X. Liu *et al.*, 2023). The mechanism of disulfide cell death involves the large-scale transport of cystine into cells through SLC7A11. Under conditions of glucose deprivation and lack of NADPH, a large accumulation of cystine and disulfides occurs within the cell, inducing a disulfide stress state, which ultimately triggers associated cell death (Goji, Takahara, Negishi, & Katoh, 2017; Joly, Delfarah, Phung, Parrish, & Graham, 2020; X. Liu *et al.*, 2023). Although caspase-8 is involved in regulating this cell death process, it is different from the protective cell death forms like caspase-8-mediated apoptosis and necroptosis under normal physiological conditions (Tummers & Green, 2017). Based on this novel form of cell death, investigating disulfide death-related genes closely associated with GBM could help deepen our understanding of its pathogenesis, which holds significant implications for clinical diagnosis, treatment, and prognosis assessment.

This study utilizes bioinformatics techniques and various machine learning algorithms to screen for biomarkers associated with GBM. These findings will help us understand the role of disulfide-related cell death in the occurrence, development, and invasion of GBM and assist in further analyzing the molecular mechanisms of GBM progression, providing guidance for clinical diagnosis and treatment.

Results

Identification of Correlative Genes of DE-DSGs

Differential analysis was performed using TCGA GBM data and GTEx normal sample data, and a volcano plot of significantly different genes was constructed (**Figure1a**). Among the 16 disulfide death-related genes, 5 genes showed significant differential expression between TCGA GBM samples and GTEx normal samples, including 2 upregulated genes and 3 downregulated genes (**Figure1b,-c**). The goal is to further investigate genes potentially associated with disulfide-related mortality, correlation analysis was performed between these 5 genes and all genes in the TCGA GBM data, followed by Lasso and univariate Cox regression analyses, resulting in 24 independent prognostic genes (**Figure1d-f**). To further explore the biological functions and potential signaling pathways associated with these 24 genes, we performed Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analyses. GO enrichment results (**Figure1g**) indicated that, in biological processes, these genes were mainly associated with T-cell regulation and protein tyrosine kinase regulation. In cellular components, they were primarily related to the endoplasmic reticulum and postsynaptic membranes. Molecular functions were primarily associated with cytokine receptors, growth factor binding, and ribonuclease activity. KEGG analysis (**Figure1h**) revealed that these genes were mainly involved in glycosaminoglycan biosynthesis, protein transport, and cytokine receptor interaction pathways. These analyses provided some insights for further investigation.

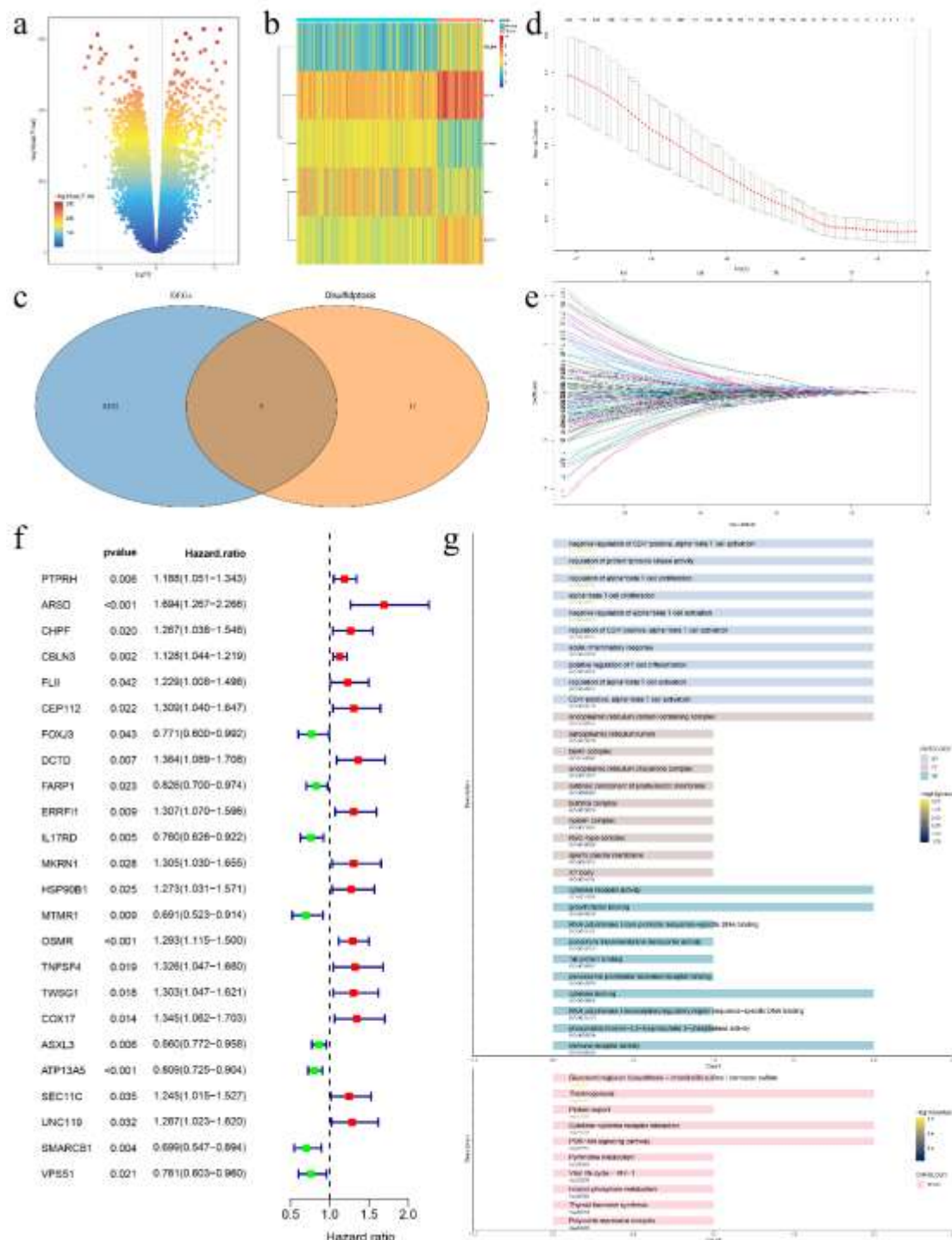


Figure1 Identification of independent prognostic genes related to disulfidptosis in glioblastoma. a. Differential analysis of TCGA sample data and volcano plot; **b-c.** Intersection of differentially expressed genes with disulfidptosis-related genes and heatmap plotting; **d-e.** Lasso regression analysis to screen for correlated genes; **f.** Univariate correlation analysis further identifying independent prognostic genes related to disulfidptosis; **g.** GO and KEGG enrichment analysis to explore the potential biological processes and signaling pathways involved in these independent prognostic genes

Identification of Twelve Correlative Genes of DE-DSGs as Diagnostic Genes for GBM

To further evaluate the diagnostic potential of these differentially expressed genes for GBM, we

applied four machine learning algorithms to identify more representative genes. Using 10-fold

cross-validation with the Lasso regression algorithm, we identified 17 candidate genes (**Figure2a-c**). After analyzing the 24 genes using the XGBoost algorithm, 23 candidate genes were selected (**Figure2d**). Random Forest algorithm analysis of the 24 genes also selected genes that met the algorithm's criteria (**Figure2e-f**). SVM-RFE algorithm analysis revealed the smallest error rate at 16 iterations (Figure2g), leading to the selection of 16 candidate genes. Finally, 12 diagnostic genes identified by all four algorithms , for example, ARSD, ASXL3, CBLN3, DCTD,

FOXJ3, IL17RD, OSMR, SEC11C, SMARCB1, TNFSF4, TWSG1, and UNC119, were selected for further analysis (**Figure2g**). Next, we constructed an ROC curve using these 12 diagnostic genes to assess the predictive performance of the SVM-RFE algorithm, which showed an AUC of 0.996 (**Figure3a**), indicating the reliability of the SVM-RFE model. Additionally, most of the AUC values were above 0.7 (**Figure3b**), suggesting that these genes have high diagnostic efficacy for GBM.

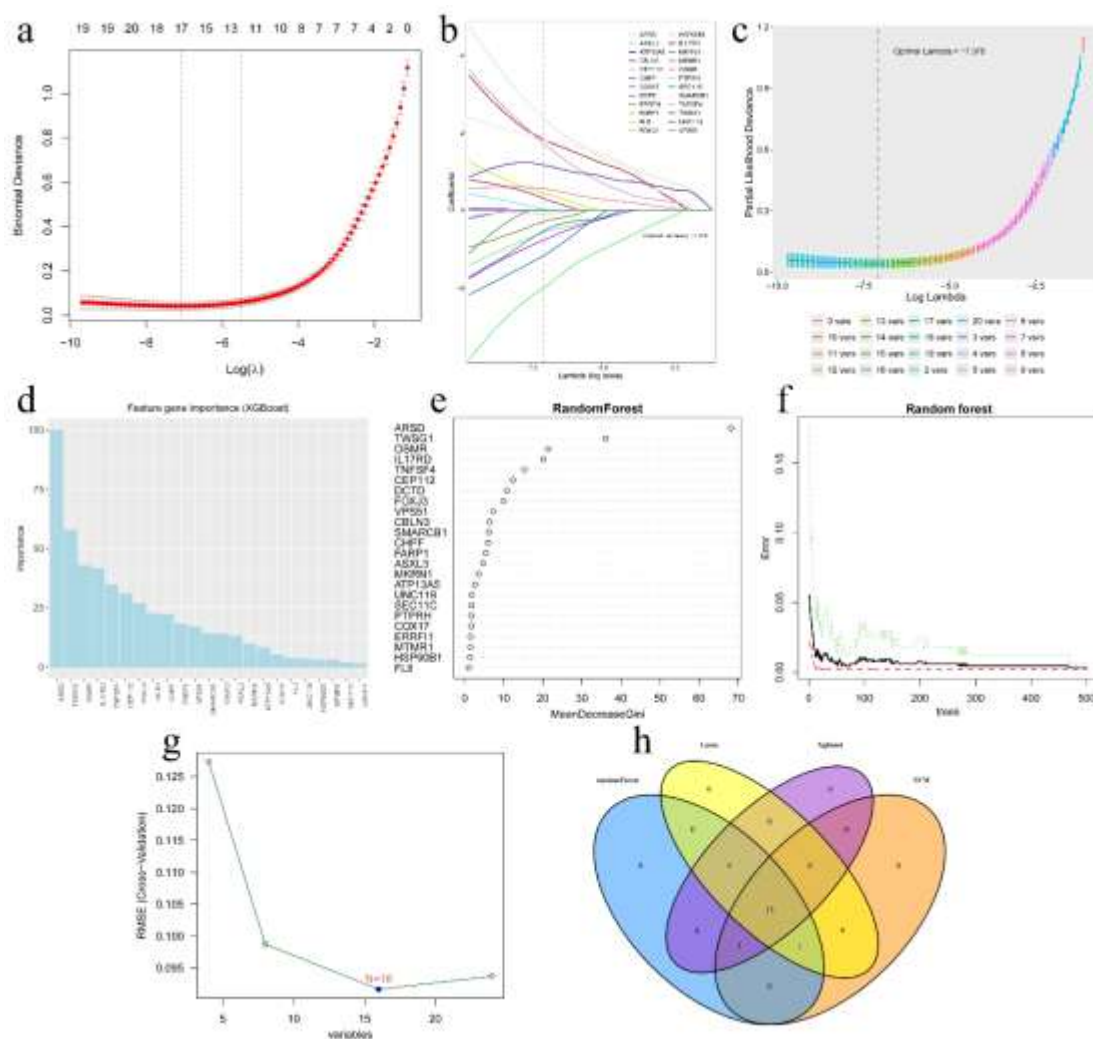


Figure2 Further screening of biomarkers in glioblastoma using four machine learning algorithms. a-c. Lasso regression analysis for screening and identifying candidate genes; d. XGBoost algorithm for screening and identifying candidate genes; e-f. Random forest algorithm for screening and identifying candidate genes; g. SVM-RFE algorithm for screening and identifying candidate genes; h. Intersection of candidate genes identified by the four machine learning algorithms

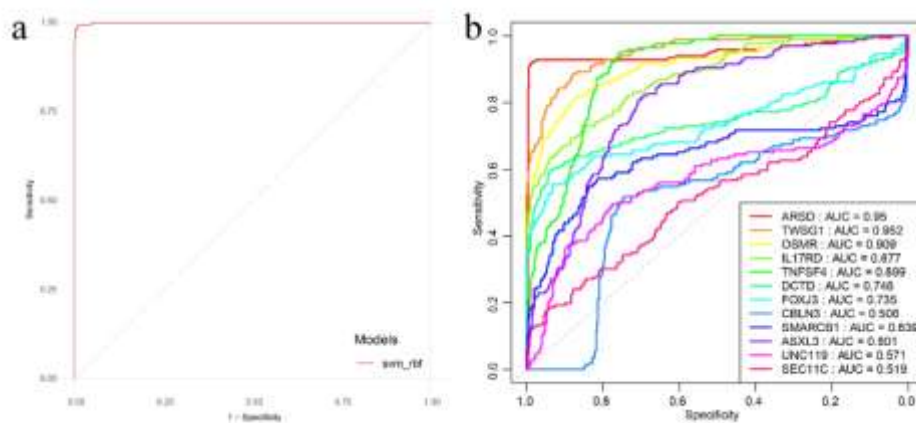


Figure3 Evaluation of machine learning algorithm performance and construction of ROC curves for each candidate gene. **a.** Model construction to evaluate the performance of the SVM-RFE algorithm; **b.** Construction of ROC curves for the 12 candidate genes to evaluate their diagnostic performance

Survival Analysis of Twelve Signature Genes

To assess the impact of these 12 signature genes on GBM prognosis and further identify more meaningful diagnostic biomarkers, we conducted survival analysis using TCGA GBM data. As shown in Figure4, high expression of DCTD ($p=0.014$, HR=1.589, **Figure4a**), TWSG1 ($p=0.009$, HR=1.606, **Figure4b**), SEC11C ($p=0.044$, HR=1.458, **Figure4d**), and OSMR ($p=0.010$, HR=1.602, **Figure4e**) was associated with poorer prognosis in GBM patients, while low expression of IL17RD ($p=0.010$, HR=0.621, **Figure4c**) and SMARCB1 ($p=0.009$, HR=0.619,

Figure4f) was linked to poorer prognosis. Additionally, ARSD ($p=0.070$, HR=1.402, **FigureS1a**), ASXL3 ($p=0.114$, HR=0.743, **FigureS1b**), CBLN3 ($p=0.119$, HR=1.336, **FigureS1c**), FOXJ3 ($p=0.097$, HR=0.734, **FigureS1d**), TNFSF4 ($p=0.582$, HR=1.11, **FigureS1e**), and UNC119 ($p=0.151$, HR=1.31, **FigureS1f**) showed no significant impact on the prognosis of GBM patients.

UNC119 ($p=0.151$, HR=1.31, **FigureS1f**) showed no significant impact on the prognosis of GBM patients.

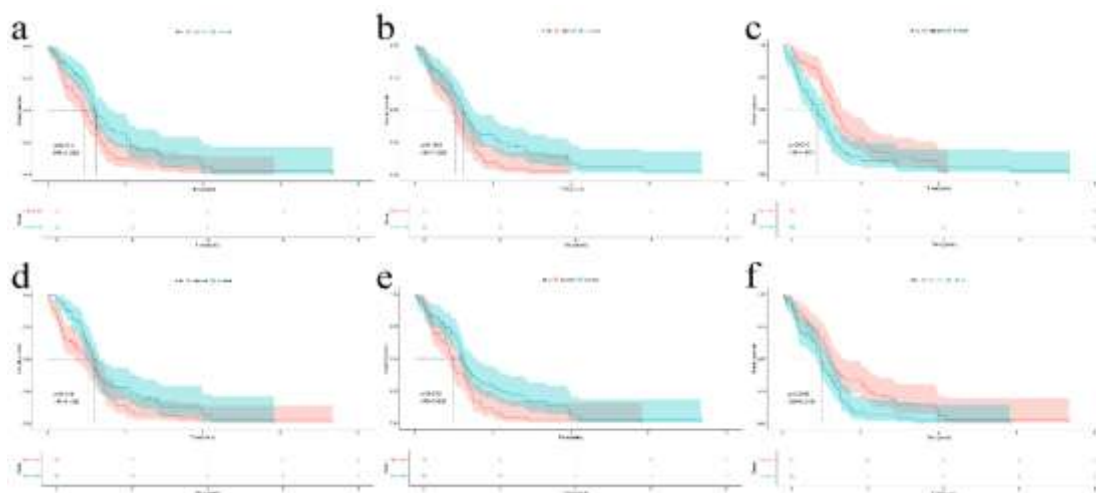


Figure4 Kaplan-Meier survival analysis of the candidate genes and their prognostic significance for glioblastoma patients. **a.** DCTD; **b.** TWSG1; **c.** IL17RD; **d.** SEC11C; **e.** OSMR; **f.** SMARCB1

Association of Signature Diagnostic Genes with Oncogenic Pathways

To further investigate the potential functional roles of these six prognostic genes, we performed GSEA pathway enrichment analysis. As shown in Figures 5a-f, DCTD, TWSG1, OSMR, SMARCB1, and IL17RD were all enriched in the CYTOKINE_JAK_STAT3 Signaling Pathway and Translation Initiation signaling pathways. SMARCB1, IL17RD, and SEC11C were enriched in the Mitochondrial Complex UCP1 in Thermogenesis pathway. Additionally, they were associated with pathways such as ITGA_B_RHOG_RAC and ITGA_B_TALIN_VINCULIN.

We then conducted GSVA enrichment analysis to evaluate pathway activation patterns between high and low expression groups of each diagnostic gene. From Figure 6, we observed that high expression levels of DCTD, OSMR, and TWSG1 were associated with activation of the IL6_JAK_STAT3 signaling pathway, interferon-

related pathways, P53 signaling pathway, NFkB_TNFA signaling pathway, and apoptosis (**Figure 6a-c**). Low expression levels of IL17RD and SMARCB1 were associated with the IL6_JAK_STAT3 signaling pathway, P53 signaling pathway, and apoptosis (**Figure 6e-f**). Furthermore, high expression levels of SEC11C were primarily associated with oxidative phosphorylation, fatty acid metabolism, P53, and KRAS signaling pathways (**Figure 6d**).

These results indicate that these six diagnostic genes are closely related to several tumor-related signaling pathways, including the JAK_STAT signaling pathway and P53 signaling pathway, both of which are crucial in tumorigenesis, progression, and diagnostic therapy. Notably, the JAK_STAT signaling pathway was highly correlated with DCTD, TWSG1, OSMR, SMARCB1, and IL17RD in both GSEA and GSVA analyses. These findings suggest that these diagnostic genes may regulate the biological processes of GBM through these signaling pathways.

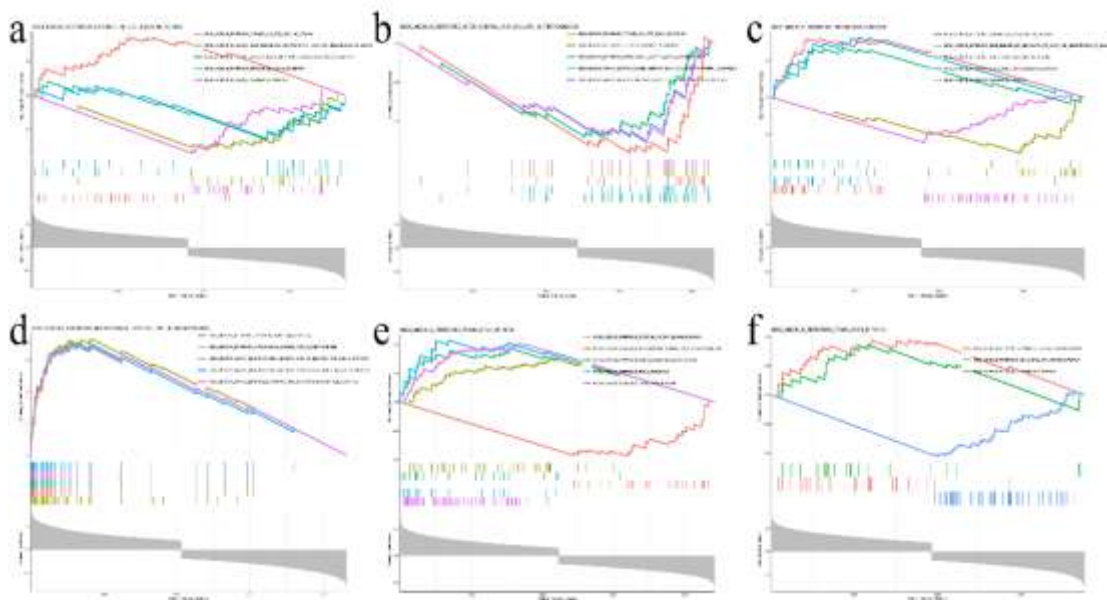


Figure 5 Single-gene GSEA signal pathway enrichment analysis of the six candidate genes with prognostic significance. a. DCTD; b. TWSG1; c. OSMR; d. SMARCB1; e. IL17RD; f. SEC11C

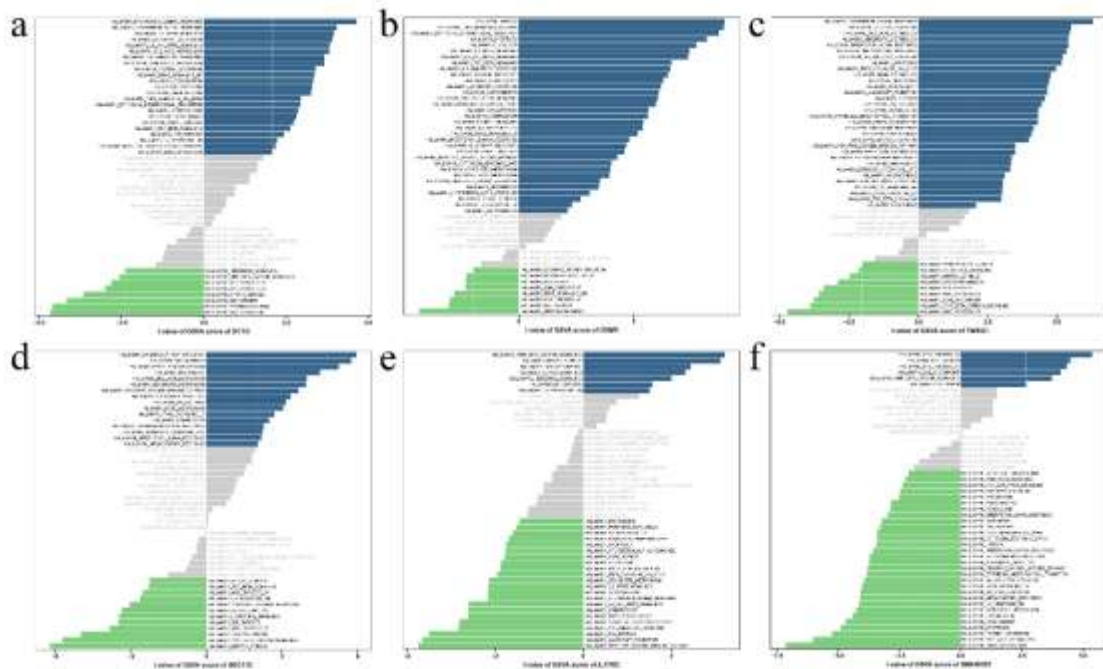


Figure6 Single-gene GSVA analysis of the six candidate genes with prognostic significance to explore their potential relationship with classical tumor-related signaling pathways. a. DCTD; b. OSMR; c. TWSG1; d. SEC11C; e. IL17RD; f. SMARCB1

Differential Analysis of Signaling Pathways and Their Correlation with Marker Genes

To further investigate whether these potential signaling pathways exhibit differential expression between GBM and normal samples and assess the correlation of these pathways with marker genes, we performed differential analysis on the GSVA scores using the limma package and evaluated their association with the marker genes. A volcano plot of differential signaling pathways between GBM and normal samples was generated, with the top five signaling pathways of high and low expression annotated (**Figure7a**). The E2F_TARGETS, G2M_CHECKPOINT, INTERFERON_GAMMA_RESPONSE, INTERFERON_ALPHA_RESPONSE, and IL6_JAK_STAT3 pathways were highly expressed in GBM, while KRAS, HEDGEHOG

signaling pathways, and oxidative phosphorylation showed a trend of low expression. In Figure7b, we found that DCTD, OSMR, and TWSG1 were positively correlated with multiple cancer-related signaling pathways, such as the IL6_JAK_STAT3 signaling pathway, P53 signaling pathway, PI3K_AKT_MTOR signaling pathway, and apoptosis signaling pathway. Conversely, IL17RD and SMARCB1 showed negative correlations with these pathways. Furthermore, SEC11C was primarily positively correlated with the P53 signaling pathway, bile acid metabolism, and oxidative phosphorylation pathways. These results further confirm previous findings, suggesting that these diagnostic genes may regulate the progression of GBM through these signaling pathways, potentially serving as therapeutic targets for GBM.

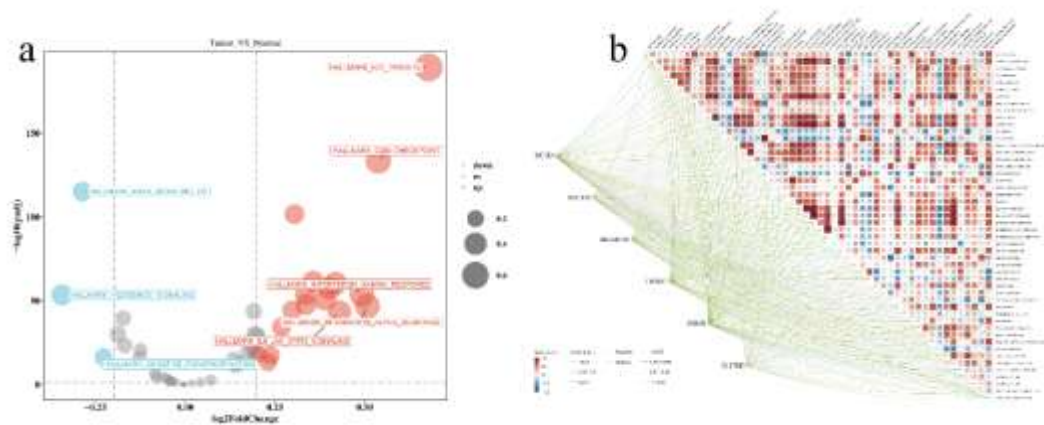


Figure7 Differential analysis of signaling pathways in glioblastoma and normal samples and the correlation between the six candidate genes and these signaling pathways. a. Analysis of differential expression of these signaling pathways in glioblastoma and normal samples with a volcano plot; b.

Heatmap of the correlation between the six candidate genes and each signaling pathway

Prediction of Targeted Therapeutics and Construction of ceRNA Networks

To further explore potential targeted drugs acting on these six genes, we used the DGIdb database to predict the target drugs for these genes,

followed by visualization of the drug-gene interactions using Cytoscape software (**Figure8a**). A total of 68 drugs were identified, with the drug prediction results shown in **Table S3**.

Table S3.

Medicine	Gene
Chlortetracycline	OSMR
0316684-0000	OSMR
Ciclopirox	OSMR
5109870	OSMR
Doxifluridine	OSMR
Tamoxifen	OSMR
Ciglitazone	OSMR
Tetradioxin	OSMR
Gefitinib	OSMR
Cadmium	OSMR
Estradiol	OSMR
Paclitaxel	OSMR
Tretinoin	OSMR
Benzo(a)pyrene	OSMR
Sarin	OSMR
Acetaminophen	OSMR
Aflatoxin B1	OSMR
Cyclosporin A	OSMR
Irinotecan hydrochloride	OSMR
Deptropine	TWSG1
Cinoxacin	TWSG1
Vorinostat	TWSG1
Scriptaid	TWSG1

Helveticoside	TWSG1
Trichostatin A	TWSG1
Arsenite	TWSG1
Copper sulfate	TWSG1
Acetaminophen	TWSG1
Cobaltous chloride	TWSG1
Aflatoxin B1	TWSG1
Rifabutin	SMARCB1
Vorinostat	SMARCB1
Trichostatin A	SMARCB1
Scriptaid	SMARCB1
Sucrose	SMARCB1
Artemisinin	SMARCB1
Dichloromethane	SMARCB1
2,2',3',4,4',5-hexachlorobiphenyl	SMARCB1
Epigallocatechin gallate	SMARCB1
Potassium chromate(VI)	SMARCB1
3,4,5,3',4'-pentachlorobiphenyl	SMARCB1
Valproic acid	SMARCB1
Sulpiride	DCTD
Chlortetracycline	DCTD
Chlorzoxazone	DCTD
Cimetidine	DCTD
Cefadroxil	DCTD
Chlorhexidine	DCTD
PHA-00665752	DCTD
Captopril	DCTD
Vorinostat	DCTD
Azacyclonol	DCTD
Dirithromycin	DCTD
Medrysone	DCTD
Ampyrone	DCTD
Clopamide	DCTD
Deptropine	DCTD
Trichostatin A	DCTD
Lobeline	DCTD
Ajmaline	DCTD
Clindamycin	DCTD
Glibenclamide	DCTD
Meclofenoxate	DCTD
Danazol	DCTD
Scriptaid	DCTD
PNU-0293363	DCTD
0175029-0000	DCTD
Gemcitavine hydrochloride	DCTD
Paclitaxel	DCTD
Cobaltous chloride	DCTD
Cyclosporin A	DCTD
Valproic Acid	DCTD
Quercetin	SEC11C
Copper	SEC11C
Hydralazine	SEC11C

Thapsigargin	SEC11C
Cycloheximide	SEC11C
Tetradioxin	SEC11C
Copper sulfate	SEC11C
Retinoic acid	SEC11C
Acetaminophen	SEC11C
Cyclosporin A	SEC11C
Valproic Acid	SEC11C
Bortezomib	IL17RD
Hydrogen peroxide	IL17RD
Estradiol	IL17RD
Mono-(2-ethylhexyl)phthalate	IL17RD
Trichostatin A	IL17RD
Valproic Acid	IL17RD
Progesterone	IL17RD

Specifically, OSMR was predicted to interact with 19 drugs, TWSG1 with 11 drugs, SMARCB1 with 12 drugs, DCTD with 30 drugs, SEC11C with 11 drugs, and IL17RD with 7 drugs.

Next, we predicted the miRNAs and lncRNAs associated with these marker genes using the miRDB, TargetScan, and starBase databases, constructing a lncRNA-miRNA-mRNA regulatory network (**Figure8b**). This network included 258 nodes, comprising the 6 marker genes, 44 miRNAs, and 208 lncRNAs. A total of 225 lncRNAs were found to competitively bind 6 miRNAs (including hsa-miR-3619-5p, hsa-miR-

214-3p) to regulate IL17RD. Among them, 52 lncRNAs competed with hsa-miR-3619-5p, and 51 lncRNAs competed with hsa-miR-214-3p. Additionally, 82 lncRNAs competed with two miRNAs (including hsa-miR-3163 and hsa-miR-340-5p) to regulate TWSG1, with 56 lncRNAs targeting hsa-miR-3163 and 26 lncRNAs targeting hsa-miR-340-5p. Moreover, 26 lncRNAs regulated DCTD by competitively binding hsa-miR-516b-5b, and 40 lncRNAs targeted hsa-miR-3118 to regulate OSMR. However, no lncRNAs were found to regulate SMARCB1 and SEC11C. Detailed interactions are listed in Table S4 and Table S5.

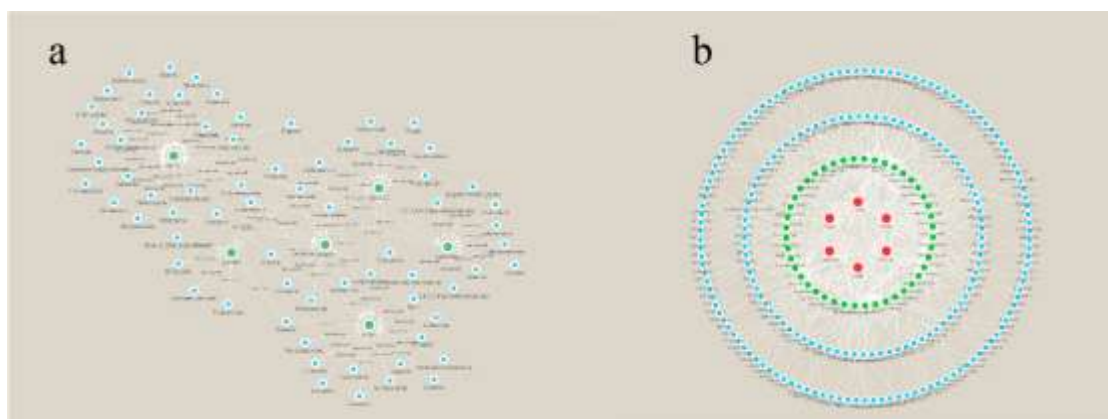


Figure8 Use of databases to predict target drugs for each gene and the construction of a network diagram of lncRNA-miRNA-mRNA interactions to explore their more complex regulatory mechanisms. a. Network diagram showing the relationship between each candidate gene and its

targeted drugs; b. Construct a network diagram of the interaction between lncRNA, miRNA, and mRNA

Table s4

Source	Target
hsa-miR-3163	TWSG1
hsa-miR-340-5p	TWSG1
hsa-miR-5694	TWSG1
hsa-miR-6128	TWSG1
hsa-miR-4282	TWSG1
hsa-miR-4528	TWSG1
hsa-miR-548t-3p	TWSG1
hsa-miR-548aa	TWSG1
hsa-miR-548ap-3p	TWSG1
hsa-miR-1299	TWSG1
hsa-miR-664a-3p	TWSG1
hsa-miR-4477a	TWSG1
MIR29B2CHG	hsa-miR-340-5p
THUMPD3-AS1	hsa-miR-340-5p
AC093010.2	hsa-miR-340-5p
SLC26A4-AS1	hsa-miR-340-5p
AC090739.1	hsa-miR-340-5p
ERICD	hsa-miR-340-5p
AL158206.1	hsa-miR-340-5p
EBLN3P	hsa-miR-340-5p
SNHG1	hsa-miR-340-5p
NEAT1	hsa-miR-340-5p
AC026356.1	hsa-miR-340-5p
TRHDE-AS1	hsa-miR-340-5p
HELLPAR	hsa-miR-340-5p
AC108704.2	hsa-miR-340-5p
SNHG14	hsa-miR-340-5p
OIP5-AS1	hsa-miR-340-5p
DNAAF4-CCPG1	hsa-miR-340-5p
AC021483.2	hsa-miR-340-5p
AC020978.7	hsa-miR-340-5p
CCDC144NL-AS1	hsa-miR-340-5p
LINC00662	hsa-miR-340-5p
AC008555.6	hsa-miR-340-5p
AC005261.1	hsa-miR-340-5p
LINC01694	hsa-miR-340-5p
XIST	hsa-miR-340-5p
LINC00630	hsa-miR-340-5p
Z99943.1	hsa-miR-3163
DNM3OS	hsa-miR-3163
TNR-IT1	hsa-miR-3163
AC012358.1	hsa-miR-3163
CHROMR	hsa-miR-3163
KANSL1L-AS1	hsa-miR-3163
AC016717.2	hsa-miR-3163
FGD5-AS1	hsa-miR-3163

AC096921.2	hsa-miR-3163
AC022336.2	hsa-miR-3163
AC093866.1	hsa-miR-3163
AC010261.2	hsa-miR-3163
AC021078.1	hsa-miR-3163
AL031779.1	hsa-miR-3163
LINC00472	hsa-miR-3163
MANEA-DT	hsa-miR-3163
AC007036.3	hsa-miR-3163
MAGI2-AS3	hsa-miR-3163
AC000123.3	hsa-miR-3163
AC004908.3	hsa-miR-3163
OTUD6B-AS1	hsa-miR-3163
ARRDC1-AS1	hsa-miR-3163
AL132656.4	hsa-miR-3163
OLMALINC	hsa-miR-3163
KCNQ1OT1	hsa-miR-3163
SNHG1	hsa-miR-3163
NEAT1	hsa-miR-3163
MALAT1	hsa-miR-3163
MIR100HG	hsa-miR-3163
AC008124.1	hsa-miR-3163
AL445223.1	hsa-miR-3163
AL356019.2	hsa-miR-3163
PSMA3-AS1	hsa-miR-3163
SNHG14	hsa-miR-3163
OIP5-AS1	hsa-miR-3163
AC051619.6	hsa-miR-3163
DNAAF4-CCPG1	hsa-miR-3163
ST20-AS1	hsa-miR-3163
CHASERR	hsa-miR-3163
AC092718.3	hsa-miR-3163
AC133540.1	hsa-miR-3163
FAM106A	hsa-miR-3163
LINC02693	hsa-miR-3163
AC125257.1	hsa-miR-3163
AC006141.1	hsa-miR-3163
AC018628.1	hsa-miR-3163
SNHG16	hsa-miR-3163
AC243964.4	hsa-miR-3163
AC005261.1	hsa-miR-3163
NORAD	hsa-miR-3163
DSCAM-AS1	hsa-miR-3163
LINC01311	hsa-miR-3163
AP000553.2	hsa-miR-3163
MIR222HG	hsa-miR-3163
XIST	hsa-miR-3163
Z83843.1	hsa-miR-3163
hsa-miR-516b-5p	DCTD
hsa-miR-1304-5p	DCTD
hsa-miR-1252-3p	DCTD
hsa-miR-892b	DCTD

AL645608.7	hsa-miR-516b-5p
AL590666.2	hsa-miR-516b-5p
AL691482.3	hsa-miR-516b-5p
PCBP1-AS1	hsa-miR-516b-5p
LINC01123	hsa-miR-516b-5p
LINC01106	hsa-miR-516b-5p
PAX8-AS1	hsa-miR-516b-5p
LINC01184	hsa-miR-516b-5p
LINC02487	hsa-miR-516b-5p
AC079781.5	hsa-miR-516b-5p
SND1-IT1	hsa-miR-516b-5p
AC016065.1	hsa-miR-516b-5p
RPARP-AS1	hsa-miR-516b-5p
H19	hsa-miR-516b-5p
KCNQ1OT1	hsa-miR-516b-5p
NEAT1	hsa-miR-516b-5p
HELLPAR	hsa-miR-516b-5p
AC131212.3	hsa-miR-516b-5p
LINC00641	hsa-miR-516b-5p
AC092718.6	hsa-miR-516b-5p
AC092127.1	hsa-miR-516b-5p
AC016876.2	hsa-miR-516b-5p
LINC00324	hsa-miR-516b-5p
TBC1D3P1-DHX40P1	hsa-miR-516b-5p
DLGAP1-AS2	hsa-miR-516b-5p
MIR23AHG	hsa-miR-516b-5p
AL133325.3	hsa-miR-516b-5p
GUSBP11	hsa-miR-516b-5p
MIAT	hsa-miR-516b-5p
hsa-miR-6809-3p	IL17RD
hsa-miR-22-3p	IL17RD
hsa-miR-4441	IL17RD
hsa-miR-6878-3p	IL17RD
hsa-miR-6736-3p	IL17RD
hsa-miR-6754-5p	IL17RD
hsa-miR-761	IL17RD
hsa-miR-3619-5p	IL17RD
hsa-miR-193a-3p	IL17RD
hsa-miR-214-3p	IL17RD
hsa-miR-4270	IL17RD
hsa-miR-193b-3p	IL17RD
hsa-let-7c-3p	IL17RD
hsa-miR-4291	IL17RD
hsa-miR-7160-5p	IL17RD
hsa-miR-3613-3p	IL17RD
MIR34AHG	hsa-miR-22-3p
LINC00339	hsa-miR-22-3p
AL360012.1	hsa-miR-22-3p
AL162431.1	hsa-miR-22-3p
AC119427.2	hsa-miR-22-3p
AC005538.2	hsa-miR-22-3p
FGD5-AS1	hsa-miR-22-3p

SH3BP5-AS1	hsa-miR-22-3p
AC079228.1	hsa-miR-22-3p
AC021078.1	hsa-miR-22-3p
SLC25A25-AS1	hsa-miR-22-3p
LINC00963	hsa-miR-22-3p
AL390294.1	hsa-miR-22-3p
H19	hsa-miR-22-3p
AC091564.7	hsa-miR-22-3p
NEAT1	hsa-miR-22-3p
MALAT1	hsa-miR-22-3p
COLCA1	hsa-miR-22-3p
SNHG14	hsa-miR-22-3p
PWAR5	hsa-miR-22-3p
AC022613.1	hsa-miR-22-3p
OIP5-AS1	hsa-miR-22-3p
Z92544.1	hsa-miR-22-3p
AC009133.1	hsa-miR-22-3p
AC135048.1	hsa-miR-22-3p
SNHG29	hsa-miR-22-3p
SNHG16	hsa-miR-22-3p
AC087741.1	hsa-miR-22-3p
NORAD	hsa-miR-22-3p
GUSBP11	hsa-miR-22-3p
AL008721.2	hsa-miR-22-3p
MIRLET7BHG	hsa-miR-22-3p
LINC00630	hsa-miR-22-3p
Z68871.1	hsa-miR-22-3p
AL162431.1	hsa-miR-193a-3p
CYTOR	hsa-miR-193a-3p
MIR4435-2HG	hsa-miR-193a-3p
AC017002.3	hsa-miR-193a-3p
LINC01184	hsa-miR-193a-3p
AL353795.3	hsa-miR-193a-3p
SNHG7	hsa-miR-193a-3p
KCNQ1OT1	hsa-miR-193a-3p
MIR194-2HG	hsa-miR-193a-3p
NEAT1	hsa-miR-193a-3p
AP001273.1	hsa-miR-193a-3p
LINC01089	hsa-miR-193a-3p
AC135050.6	hsa-miR-193a-3p
AC010327.6	hsa-miR-193a-3p
ZFAS1	hsa-miR-193a-3p
MIRLET7BHG	hsa-miR-193a-3p
XIST	hsa-miR-193a-3p
AL162431.1	hsa-miR-193b-3p
CYTOR	hsa-miR-193b-3p
MIR4435-2HG	hsa-miR-193b-3p
AC017002.3	hsa-miR-193b-3p
LINC01184	hsa-miR-193b-3p
AL353795.3	hsa-miR-193b-3p
SNHG7	hsa-miR-193b-3p
KCNQ1OT1	hsa-miR-193b-3p

MIR194-2HG	hsa-miR-193b-3p
NEAT1	hsa-miR-193b-3p
AP001273.1	hsa-miR-193b-3p
LINC01089	hsa-miR-193b-3p
AC135050.6	hsa-miR-193b-3p
AC092295.2	hsa-miR-193b-3p
AC010327.6	hsa-miR-193b-3p
ZFAS1	hsa-miR-193b-3p
MIRLET7BHG	hsa-miR-193b-3p
XIST	hsa-miR-193b-3p
AL645608.2	hsa-miR-214-3p
AL031282.2	hsa-miR-214-3p
MIR34AHG	hsa-miR-214-3p
LINC00339	hsa-miR-214-3p
SLFNL1-AS1	hsa-miR-214-3p
LINC01140	hsa-miR-214-3p
MIR29B2CHG	hsa-miR-214-3p
AL160408.3	hsa-miR-214-3p
AC012513.3	hsa-miR-214-3p
LINC02832	hsa-miR-214-3p
FGD5-AS1	hsa-miR-214-3p
TMCC1-AS1	hsa-miR-214-3p
PP7080	hsa-miR-214-3p
SLC9A3-AS1	hsa-miR-214-3p
TMEM161B-AS1	hsa-miR-214-3p
EPB41L4A-AS1	hsa-miR-214-3p
CASC15	hsa-miR-214-3p
CT66	hsa-miR-214-3p
STAG3L5P-PVRIG2P-PILRB	hsa-miR-214-3p
AC069281.2	hsa-miR-214-3p
AP003472.2	hsa-miR-214-3p
AC084125.2	hsa-miR-214-3p
AL157392.3	hsa-miR-214-3p
AP006621.5	hsa-miR-214-3p
KCNQ1OT1	hsa-miR-214-3p
NEAT1	hsa-miR-214-3p
AP003119.3	hsa-miR-214-3p
AC026362.1	hsa-miR-214-3p
MIR17HG	hsa-miR-214-3p
LINC00641	hsa-miR-214-3p
Z98883.1	hsa-miR-214-3p
AC106820.4	hsa-miR-214-3p
AC020663.4	hsa-miR-214-3p
AC109460.3	hsa-miR-214-3p
SLX1A-SULT1A3	hsa-miR-214-3p
AC113189.4	hsa-miR-214-3p
AC055811.4	hsa-miR-214-3p
SNHG16	hsa-miR-214-3p
AC087741.1	hsa-miR-214-3p
LINC00482	hsa-miR-214-3p
LINC01535	hsa-miR-214-3p
AC093227.1	hsa-miR-214-3p

PCAT19	hsa-miR-214-3p
A1BG-AS1	hsa-miR-214-3p
AL121906.2	hsa-miR-214-3p
SNHG17	hsa-miR-214-3p
TSPEAR-AS2	hsa-miR-214-3p
AC002470.2	hsa-miR-214-3p
MIRLET7BHG	hsa-miR-214-3p
C22orf34	hsa-miR-214-3p
XIST	hsa-miR-214-3p
AL645608.2	hsa-miR-761
AL031282.2	hsa-miR-761
MIR34AHG	hsa-miR-761
LINC00339	hsa-miR-761
SLFNL1-AS1	hsa-miR-761
LINC01140	hsa-miR-761
MIR29B2CHG	hsa-miR-761
AL160408.3	hsa-miR-761
AC012513.3	hsa-miR-761
LINC02832	hsa-miR-761
FGD5-AS1	hsa-miR-761
TMCC1-AS1	hsa-miR-761
PP7080	hsa-miR-761
SLC9A3-AS1	hsa-miR-761
TMEM161B-AS1	hsa-miR-761
EPB41L4A-AS1	hsa-miR-761
CASC15	hsa-miR-761
CT66	hsa-miR-761
STAG3L5P-PVRIG2P-PILRB	hsa-miR-761
AC069281.2	hsa-miR-761
AP003472.2	hsa-miR-761
AC084125.2	hsa-miR-761
AL157392.3	hsa-miR-761
SGMS1-AS1	hsa-miR-761
AP006621.5	hsa-miR-761
KCNQ1OT1	hsa-miR-761
NEAT1	hsa-miR-761
AP003119.3	hsa-miR-761
AC026362.1	hsa-miR-761
MIR17HG	hsa-miR-761
LINC00641	hsa-miR-761
Z98883.1	hsa-miR-761
AC106820.4	hsa-miR-761
AC020663.4	hsa-miR-761
AC109460.3	hsa-miR-761
SLX1A-SULT1A3	hsa-miR-761
AC113189.4	hsa-miR-761
AC055811.4	hsa-miR-761
SNHG16	hsa-miR-761
AC087741.1	hsa-miR-761
LINC00482	hsa-miR-761
LINC01535	hsa-miR-761
AC093227.1	hsa-miR-761

PCAT19	hsa-miR-761
AC073548.1	hsa-miR-761
A1BG-AS1	hsa-miR-761
AL121906.2	hsa-miR-761
SNHG17	hsa-miR-761
TSPEAR-AS2	hsa-miR-761
AC002470.2	hsa-miR-761
MIRLET7BHG	hsa-miR-761
C22orf34	hsa-miR-761
XIST	hsa-miR-761
AL645608.2	hsa-miR-3619-5p
AL031282.2	hsa-miR-3619-5p
MIR34AHG	hsa-miR-3619-5p
LINC00339	hsa-miR-3619-5p
SLFNL1-AS1	hsa-miR-3619-5p
LINC01140	hsa-miR-3619-5p
MIR29B2CHG	hsa-miR-3619-5p
AL160408.3	hsa-miR-3619-5p
AC012513.3	hsa-miR-3619-5p
LINC02832	hsa-miR-3619-5p
FGD5-AS1	hsa-miR-3619-5p
TMCC1-AS1	hsa-miR-3619-5p
PP7080	hsa-miR-3619-5p
SLC9A3-AS1	hsa-miR-3619-5p
TMEM161B-AS1	hsa-miR-3619-5p
EPB41L4A-AS1	hsa-miR-3619-5p
CASC15	hsa-miR-3619-5p
CT66	hsa-miR-3619-5p
STAG3L5P-PVRIG2P-PILRB	hsa-miR-3619-5p
AC069281.2	hsa-miR-3619-5p
AP003472.2	hsa-miR-3619-5p
AC084125.2	hsa-miR-3619-5p
AL157392.3	hsa-miR-3619-5p
SGMS1-AS1	hsa-miR-3619-5p
KCNQ1OT1	hsa-miR-3619-5p
NEAT1	hsa-miR-3619-5p
AP003119.3	hsa-miR-3619-5p
AC026362.1	hsa-miR-3619-5p
MIR17HG	hsa-miR-3619-5p
LINC00641	hsa-miR-3619-5p
Z98883.1	hsa-miR-3619-5p
AC106820.4	hsa-miR-3619-5p
AC020663.4	hsa-miR-3619-5p
AC109460.3	hsa-miR-3619-5p
SLX1A-SULT1A3	hsa-miR-3619-5p
AC113189.4	hsa-miR-3619-5p
AC055811.4	hsa-miR-3619-5p
SNHG16	hsa-miR-3619-5p
AC087741.1	hsa-miR-3619-5p
LINC00482	hsa-miR-3619-5p
LINC01535	hsa-miR-3619-5p
AC093227.1	hsa-miR-3619-5p

PCAT19	hsa-miR-3619-5p
AC073548.1	hsa-miR-3619-5p
A1BG-AS1	hsa-miR-3619-5p
AL121906.2	hsa-miR-3619-5p
SNHG17	hsa-miR-3619-5p
TSPEAR-AS2	hsa-miR-3619-5p
AC002470.2	hsa-miR-3619-5p
MIRLET7BHG	hsa-miR-3619-5p
C22orf34	hsa-miR-3619-5p
XIST	hsa-miR-3619-5p
hsa-miR-4504	OSMR
hsa-miR-518c-5p	OSMR
hsa-miR-3118	OSMR
hsa-miR-100-3p	OSMR
hsa-miR-4501	OSMR
PINK1-AS	hsa-miR-3118
AL445231.1	hsa-miR-3118
ADAMTSL4-AS1	hsa-miR-3118
DNM3OS	hsa-miR-3118
HK2-DT	hsa-miR-3118
LINC01123	hsa-miR-3118
LINC01106	hsa-miR-3118
AC009948.1	hsa-miR-3118
LINC01963	hsa-miR-3118
AC104447.1	hsa-miR-3118
LINC02035	hsa-miR-3118
LINC01618	hsa-miR-3118
PP7080	hsa-miR-3118
SND1-IT1	hsa-miR-3118
AC007938.3	hsa-miR-3118
CDKN2B-AS1	hsa-miR-3118
AL353795.3	hsa-miR-3118
MIR600HG	hsa-miR-3118
OLMALINC	hsa-miR-3118
KCNQ1OT1	hsa-miR-3118
NEAT1	hsa-miR-3118
AC005831.1	hsa-miR-3118
AC144548.1	hsa-miR-3118
AC073592.5	hsa-miR-3118
Z92544.1	hsa-miR-3118
MIR22HG	hsa-miR-3118
AC015799.1	hsa-miR-3118
AC127521.1	hsa-miR-3118
LINC02693	hsa-miR-3118
AC005562.1	hsa-miR-3118
AC004687.1	hsa-miR-3118
AC005332.6	hsa-miR-3118
AC145207.5	hsa-miR-3118
TYMSOS	hsa-miR-3118
WDR7-OT1	hsa-miR-3118
MIR122HG	hsa-miR-3118
AC007191.1	hsa-miR-3118

AL117335.1	hsa-miR-3118
LINC01278	hsa-miR-3118
FTX	hsa-miR-3118
hsa-miR-4520-2-3p	SEC11C
hsa-miR-1185-1-3p	SEC11C
hsa-miR-1185-2-3p	SEC11C
hsa-let-7f-2-3p	SEC11C
hsa-miR-372-5p	SEC11C
hsa-miR-6832-3p	SEC11C
hsa-miR-4283	SMARCB1

Table S5

Source	Type
DCTD	mRNA-DCTD
hsa-miR-516b-5p	miRNA-DCTD
hsa-miR-1304-5p	miRNA-DCTD
hsa-miR-1252-3p	miRNA-DCTD
hsa-miR-892b	miRNA-DCTD
AL645608.7	lncRNA-DCTD
AL590666.2	lncRNA-DCTD
AL691482.3	lncRNA-DCTD
PCBP1-AS1	lncRNA-DCTD
LINC01123	lncRNA-DCTD
LINC01106	lncRNA-DCTD
PAX8-AS1	lncRNA-DCTD
LINC01184	lncRNA-DCTD
LINC02487	lncRNA-DCTD
AC079781.5	lncRNA-DCTD
SND1-IT1	lncRNA-DCTD
AC016065.1	lncRNA-DCTD
RPARP-AS1	lncRNA-DCTD
H19	lncRNA-DCTD
KCNQ1OT1	lncRNA-DCTD
NEAT1	lncRNA-DCTD
HELLPAR	lncRNA-DCTD
AC131212.3	lncRNA-DCTD
LINC00641	lncRNA-DCTD
AC092718.6	lncRNA-DCTD
AC092127.1	lncRNA-DCTD
AC016876.2	lncRNA-DCTD
LINC00324	lncRNA-DCTD
TBC1D3P1-DHX40P1	lncRNA-DCTD
DLGAP1-AS2	lncRNA-DCTD
MIR23AHG	lncRNA-DCTD
AL133325.3	lncRNA-DCTD
GUSBP11	lncRNA-DCTD
MIAT	lncRNA-DCTD
IL17RD	mRNA-IL17RD
hsa-miR-6809-3p	miRNA-IL17RD
hsa-miR-22-3p	miRNA-IL17RD
hsa-miR-4441	miRNA-IL17RD

hsa-miR-6878-3p	miRNA-IL17RD
hsa-miR-6736-3p	miRNA-IL17RD
hsa-miR-6754-5p	miRNA-IL17RD
hsa-miR-761	miRNA-IL17RD
hsa-miR-3619-5p	miRNA-IL17RD
hsa-miR-193a-3p	miRNA-IL17RD
hsa-miR-214-3p	miRNA-IL17RD
hsa-miR-4270	miRNA-IL17RD
hsa-miR-193b-3p	miRNA-IL17RD
hsa-let-7c-3p	miRNA-IL17RD
hsa-miR-4291	miRNA-IL17RD
hsa-miR-7160-5p	miRNA-IL17RD
hsa-miR-3613-3p	miRNA-IL17RD
MIR34AHG	lncRNA-IL17RD
LINC00339	lncRNA-IL17RD
AL360012.1	lncRNA-IL17RD
AL162431.1	lncRNA-IL17RD
AC119427.2	lncRNA-IL17RD
AC005538.2	lncRNA-IL17RD
FGD5-AS1	lncRNA-IL17RD
SH3BP5-AS1	lncRNA-IL17RD
AC079228.1	lncRNA-IL17RD
AC021078.1	lncRNA-IL17RD
SLC25A25-AS1	lncRNA-IL17RD
LINC00963	lncRNA-IL17RD
AL390294.1	lncRNA-IL17RD
H19	lncRNA-IL17RD
AC091564.7	lncRNA-IL17RD
NEAT1	lncRNA-IL17RD
MALAT1	lncRNA-IL17RD
COLCA1	lncRNA-IL17RD
SNHG14	lncRNA-IL17RD
PWAR5	lncRNA-IL17RD
AC022613.1	lncRNA-IL17RD
OIP5-AS1	lncRNA-IL17RD
Z92544.1	lncRNA-IL17RD
AC009133.1	lncRNA-IL17RD
AC135048.1	lncRNA-IL17RD
SNHG29	lncRNA-IL17RD
SNHG16	lncRNA-IL17RD
AC087741.1	lncRNA-IL17RD
NORAD	lncRNA-IL17RD
GUSBP11	lncRNA-IL17RD
AL008721.2	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
LINC00630	lncRNA-IL17RD
Z68871.1	lncRNA-IL17RD
AL162431.1	lncRNA-IL17RD
CYTOR	lncRNA-IL17RD
MIR4435-2HG	lncRNA-IL17RD
AC017002.3	lncRNA-IL17RD
LINC01184	lncRNA-IL17RD

AL353795.3	lncRNA-IL17RD
SNHG7	lncRNA-IL17RD
KCNQ1OT1	lncRNA-IL17RD
MIR194-2HG	lncRNA-IL17RD
NEAT1	lncRNA-IL17RD
AP001273.1	lncRNA-IL17RD
LINC01089	lncRNA-IL17RD
AC135050.6	lncRNA-IL17RD
AC010327.6	lncRNA-IL17RD
ZFAS1	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
XIST	lncRNA-IL17RD
AL162431.1	lncRNA-IL17RD
CYTOR	lncRNA-IL17RD
MIR4435-2HG	lncRNA-IL17RD
AC017002.3	lncRNA-IL17RD
LINC01184	lncRNA-IL17RD
AL353795.3	lncRNA-IL17RD
SNHG7	lncRNA-IL17RD
KCNQ1OT1	lncRNA-IL17RD
MIR194-2HG	lncRNA-IL17RD
NEAT1	lncRNA-IL17RD
AP001273.1	lncRNA-IL17RD
LINC01089	lncRNA-IL17RD
AC135050.6	lncRNA-IL17RD
AC092295.2	lncRNA-IL17RD
AC010327.6	lncRNA-IL17RD
ZFAS1	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
XIST	lncRNA-IL17RD
AL645608.2	lncRNA-IL17RD
AL031282.2	lncRNA-IL17RD
MIR34AHG	lncRNA-IL17RD
LINC00339	lncRNA-IL17RD
SLFNL1-AS1	lncRNA-IL17RD
LINC01140	lncRNA-IL17RD
MIR29B2CHG	lncRNA-IL17RD
AL160408.3	lncRNA-IL17RD
AC012513.3	lncRNA-IL17RD
LINC02832	lncRNA-IL17RD
FGD5-AS1	lncRNA-IL17RD
TMCC1-AS1	lncRNA-IL17RD
PP7080	lncRNA-IL17RD
SLC9A3-AS1	lncRNA-IL17RD
TMEM161B-AS1	lncRNA-IL17RD
EPB41L4A-AS1	lncRNA-IL17RD
CASC15	lncRNA-IL17RD
CT66	lncRNA-IL17RD
STAG3L5P-PVRIG2P-PILRB	lncRNA-IL17RD
AC069281.2	lncRNA-IL17RD
AP003472.2	lncRNA-IL17RD
AC084125.2	lncRNA-IL17RD

AL157392.3	lncRNA-IL17RD
AP006621.5	lncRNA-IL17RD
KCNQ1OT1	lncRNA-IL17RD
NEAT1	lncRNA-IL17RD
AP003119.3	lncRNA-IL17RD
AC026362.1	lncRNA-IL17RD
MIR17HG	lncRNA-IL17RD
LINC00641	lncRNA-IL17RD
Z98883.1	lncRNA-IL17RD
AC106820.4	lncRNA-IL17RD
AC020663.4	lncRNA-IL17RD
AC109460.3	lncRNA-IL17RD
SLX1A-SULT1A3	lncRNA-IL17RD
AC113189.4	lncRNA-IL17RD
AC055811.4	lncRNA-IL17RD
SNHG16	lncRNA-IL17RD
AC087741.1	lncRNA-IL17RD
LINC00482	lncRNA-IL17RD
LINC01535	lncRNA-IL17RD
AC093227.1	lncRNA-IL17RD
PCAT19	lncRNA-IL17RD
A1BG-AS1	lncRNA-IL17RD
AL121906.2	lncRNA-IL17RD
SNHG17	lncRNA-IL17RD
TSPEAR-AS2	lncRNA-IL17RD
AC002470.2	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
C22orf34	lncRNA-IL17RD
XIST	lncRNA-IL17RD
AL645608.2	lncRNA-IL17RD
AL031282.2	lncRNA-IL17RD
MIR34AHG	lncRNA-IL17RD
LINC00339	lncRNA-IL17RD
SLFNL1-AS1	lncRNA-IL17RD
LINC01140	lncRNA-IL17RD
MIR29B2CHG	lncRNA-IL17RD
AL160408.3	lncRNA-IL17RD
AC012513.3	lncRNA-IL17RD
LINC02832	lncRNA-IL17RD
FGD5-AS1	lncRNA-IL17RD
TMCC1-AS1	lncRNA-IL17RD
PP7080	lncRNA-IL17RD
SLC9A3-AS1	lncRNA-IL17RD
TMEM161B-AS1	lncRNA-IL17RD
EPB41L4A-AS1	lncRNA-IL17RD
CASC15	lncRNA-IL17RD
CT66	lncRNA-IL17RD
STAG3L5P-PVRIG2P-PILRB	lncRNA-IL17RD
AC069281.2	lncRNA-IL17RD
AP003472.2	lncRNA-IL17RD
AC084125.2	lncRNA-IL17RD
AL157392.3	lncRNA-IL17RD

SGMS1-AS1	lncRNA-IL17RD
AP006621.5	lncRNA-IL17RD
KCNQ1OT1	lncRNA-IL17RD
NEAT1	lncRNA-IL17RD
AP003119.3	lncRNA-IL17RD
AC026362.1	lncRNA-IL17RD
MIR17HG	lncRNA-IL17RD
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Z98883.1	lncRNA-IL17RD
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AC020663.4	lncRNA-IL17RD
AC109460.3	lncRNA-IL17RD
SLX1A-SULT1A3	lncRNA-IL17RD
AC113189.4	lncRNA-IL17RD
AC055811.4	lncRNA-IL17RD
SNHG16	lncRNA-IL17RD
AC087741.1	lncRNA-IL17RD
LINC00482	lncRNA-IL17RD
LINC01535	lncRNA-IL17RD
AC093227.1	lncRNA-IL17RD
PCAT19	lncRNA-IL17RD
AC073548.1	lncRNA-IL17RD
A1BG-AS1	lncRNA-IL17RD
AL121906.2	lncRNA-IL17RD
SNHG17	lncRNA-IL17RD
TSPEAR-AS2	lncRNA-IL17RD
AC002470.2	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
C22orf34	lncRNA-IL17RD
XIST	lncRNA-IL17RD
AL645608.2	lncRNA-IL17RD
AL031282.2	lncRNA-IL17RD
MIR34AHG	lncRNA-IL17RD
LINC00339	lncRNA-IL17RD
SLFNL1-AS1	lncRNA-IL17RD
LINC01140	lncRNA-IL17RD
MIR29B2CHG	lncRNA-IL17RD
AL160408.3	lncRNA-IL17RD
AC012513.3	lncRNA-IL17RD
LINC02832	lncRNA-IL17RD
FGD5-AS1	lncRNA-IL17RD
TMCC1-AS1	lncRNA-IL17RD
PP7080	lncRNA-IL17RD
SLC9A3-AS1	lncRNA-IL17RD
TMEM161B-AS1	lncRNA-IL17RD
EPB41L4A-AS1	lncRNA-IL17RD
CASC15	lncRNA-IL17RD
CT66	lncRNA-IL17RD
STAG3L5P-PVRIG2P-PILRB	lncRNA-IL17RD
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AP003472.2	lncRNA-IL17RD
AC084125.2	lncRNA-IL17RD

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KCNQ1OT1	lncRNA-IL17RD
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AC020663.4	lncRNA-IL17RD
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SLX1A-SULT1A3	lncRNA-IL17RD
AC113189.4	lncRNA-IL17RD
AC055811.4	lncRNA-IL17RD
SNHG16	lncRNA-IL17RD
AC087741.1	lncRNA-IL17RD
LINC00482	lncRNA-IL17RD
LINC01535	lncRNA-IL17RD
AC093227.1	lncRNA-IL17RD
PCAT19	lncRNA-IL17RD
AC073548.1	lncRNA-IL17RD
A1BG-AS1	lncRNA-IL17RD
AL121906.2	lncRNA-IL17RD
SNHG17	lncRNA-IL17RD
TSPEAR-AS2	lncRNA-IL17RD
AC002470.2	lncRNA-IL17RD
MIRLET7BHG	lncRNA-IL17RD
C22orf34	lncRNA-IL17RD
XIST	lncRNA-IL17RD
OSMR	mRNA-OSMR
hsa-miR-4504	miRNA-OSMR
hsa-miR-518c-5p	miRNA-OSMR
hsa-miR-3118	miRNA-OSMR
hsa-miR-100-3p	miRNA-OSMR
hsa-miR-4501	miRNA-OSMR
PINK1-AS	lncRNA-OSMR
AL445231.1	lncRNA-OSMR
ADAMTSL4-AS1	lncRNA-OSMR
DNM3OS	lncRNA-OSMR
HK2-DT	lncRNA-OSMR
LINC01123	lncRNA-OSMR
LINC01106	lncRNA-OSMR
AC009948.1	lncRNA-OSMR
LINC01963	lncRNA-OSMR
AC104447.1	lncRNA-OSMR
LINC02035	lncRNA-OSMR
LINC01618	lncRNA-OSMR
PP7080	lncRNA-OSMR
SND1-IT1	lncRNA-OSMR
AC007938.3	lncRNA-OSMR
CDKN2B-AS1	lncRNA-OSMR

AL353795.3	lncRNA-OSMR
MIR600HG	lncRNA-OSMR
OLMALINC	lncRNA-OSMR
KCNQ1OT1	lncRNA-OSMR
NEAT1	lncRNA-OSMR
AC005831.1	lncRNA-OSMR
AC144548.1	lncRNA-OSMR
AC073592.5	lncRNA-OSMR
Z92544.1	lncRNA-OSMR
MIR22HG	lncRNA-OSMR
AC015799.1	lncRNA-OSMR
AC127521.1	lncRNA-OSMR
LINC02693	lncRNA-OSMR
AC005562.1	lncRNA-OSMR
AC004687.1	lncRNA-OSMR
AC005332.6	lncRNA-OSMR
AC145207.5	lncRNA-OSMR
TYMSOS	lncRNA-OSMR
WDR7-OT1	lncRNA-OSMR
MIR122HG	lncRNA-OSMR
AC007191.1	lncRNA-OSMR
AL117335.1	lncRNA-OSMR
LINC01278	lncRNA-OSMR
FTX	lncRNA-OSMR
SEC11C	mRNA-SEC11C
hsa-miR-4520-2-3p	miRNA-SEC11C
hsa-miR-1185-1-3p	miRNA-SEC11C
hsa-miR-1185-2-3p	miRNA-SEC11C
hsa-let-7f-2-3p	miRNA-SEC11C
hsa-miR-372-5p	miRNA-SEC11C
hsa-miR-6832-3p	miRNA-SEC11C
SMARCB1	mRNA-SMARCB1
hsa-miR-4283	miRNA-SMARCB1
TWSG1	mRNA-TWSG1
hsa-miR-3163	miRNA-TWSG1
hsa-miR-340-5p	miRNA-TWSG1
hsa-miR-5694	miRNA-TWSG1
hsa-miR-6128	miRNA-TWSG1
hsa-miR-4282	miRNA-TWSG1
hsa-miR-4528	miRNA-TWSG1
hsa-miR-548t-3p	miRNA-TWSG1
hsa-miR-548aa	miRNA-TWSG1
hsa-miR-548ap-3p	miRNA-TWSG1
hsa-miR-1299	miRNA-TWSG1
hsa-miR-664a-3p	miRNA-TWSG1
hsa-miR-4477a	miRNA-TWSG1
MIR29B2CHG	lncRNA-TWSG1
THUMPD3-AS1	lncRNA-TWSG1
AC093010.2	lncRNA-TWSG1
SLC26A4-AS1	lncRNA-TWSG1
AC090739.1	lncRNA-TWSG1
ERICD	lncRNA-TWSG1

AL158206.1	lncRNA-TWSG1
EBLN3P	lncRNA-TWSG1
SNHG1	lncRNA-TWSG1
NEAT1	lncRNA-TWSG1
AC026356.1	lncRNA-TWSG1
TRHDE-AS1	lncRNA-TWSG1
HELLPAR	lncRNA-TWSG1
AC108704.2	lncRNA-TWSG1
SNHG14	lncRNA-TWSG1
OIP5-AS1	lncRNA-TWSG1
DNAAF4-CCPG1	lncRNA-TWSG1
AC021483.2	lncRNA-TWSG1
AC020978.7	lncRNA-TWSG1
CCDC144NL-AS1	lncRNA-TWSG1
LINC00662	lncRNA-TWSG1
AC008555.6	lncRNA-TWSG1
AC005261.1	lncRNA-TWSG1
LINC01694	lncRNA-TWSG1
XIST	lncRNA-TWSG1
LINC00630	lncRNA-TWSG1
Z99943.1	lncRNA-TWSG1
DNM3OS	lncRNA-TWSG1
TNR-IT1	lncRNA-TWSG1
AC012358.1	lncRNA-TWSG1
CHROMR	lncRNA-TWSG1
KANSL1L-AS1	lncRNA-TWSG1
AC016717.2	lncRNA-TWSG1
FGD5-AS1	lncRNA-TWSG1
AC096921.2	lncRNA-TWSG1
AC022336.2	lncRNA-TWSG1
AC093866.1	lncRNA-TWSG1
AC010261.2	lncRNA-TWSG1
AC021078.1	lncRNA-TWSG1
AL031779.1	lncRNA-TWSG1
LINC00472	lncRNA-TWSG1
MANEA-DT	lncRNA-TWSG1
AC007036.3	lncRNA-TWSG1
MAGI2-AS3	lncRNA-TWSG1
AC000123.3	lncRNA-TWSG1
AC004908.3	lncRNA-TWSG1
OTUD6B-AS1	lncRNA-TWSG1
ARRDC1-AS1	lncRNA-TWSG1
AL132656.4	lncRNA-TWSG1
OLMALINC	lncRNA-TWSG1
KCNQ1OT1	lncRNA-TWSG1
SNHG1	lncRNA-TWSG1
NEAT1	lncRNA-TWSG1
MALAT1	lncRNA-TWSG1
MIR100HG	lncRNA-TWSG1
AC008124.1	lncRNA-TWSG1
AL445223.1	lncRNA-TWSG1
AL356019.2	lncRNA-TWSG1

PSMA3-AS1	lncRNA-TWSG1
SNHG14	lncRNA-TWSG1
OIP5-AS1	lncRNA-TWSG1
AC051619.6	lncRNA-TWSG1
DNAAF4-CCPG1	lncRNA-TWSG1
ST20-AS1	lncRNA-TWSG1
CHASERR	lncRNA-TWSG1
AC092718.3	lncRNA-TWSG1
AC133540.1	lncRNA-TWSG1
FAM106A	lncRNA-TWSG1
LINC02693	lncRNA-TWSG1
AC125257.1	lncRNA-TWSG1
AC006141.1	lncRNA-TWSG1
AC018628.1	lncRNA-TWSG1
SNHG16	lncRNA-TWSG1
AC243964.4	lncRNA-TWSG1
AC005261.1	lncRNA-TWSG1
NORAD	lncRNA-TWSG1
DSCAM-AS1	lncRNA-TWSG1
LINC01311	lncRNA-TWSG1
AP000553.2	lncRNA-TWSG1
MIR222HG	lncRNA-TWSG1
XIST	lncRNA-TWSG1
Z83843.1	lncRNA-TWSG1

Validation Using the Independent Validation Cohort

Finally, we used the GSE4290 dataset to validate the expression of these 6 marker genes in GBM and normal samples. Additionally, we used the GSE43378 and GSE83300 datasets to validate the survival analysis of these 6 marker genes in GBM patients. In the TCGA and GTEx merged data, we found that DCTD, SMARCB1, TWSG1, OSMR, and IL17RD were all highly expressed in GBM, while SEC11C showed no significant correlation (**Figure9a**). In the GSE4290 dataset, all 6 marker

genes exhibited a trend of high expression in GBM (**Figure9b**), which was consistent with the test set results. The survival analysis results in the GSE43378 and GSE83300 validation datasets showed that, except for DCTD ($p=0.101$, $HR=1.424$, **Figure9c**) and SEC11C ($p=0.232$, $HR=0.773$, **Figure9d**), which had no significant correlation, IL17RD ($p=0.030$, $HR=0.63$, **Figure9e**), TWSG1 ($p=0.003$, $HR=1.895$, **Figure9f**), OSMR ($p=0.017$, $HR=1.666$, **Figure9g**), and SMARCB1 ($p<0.001$, $HR=0.46$, **Figure9h**) had significant effects on the prognosis of GBM patients.

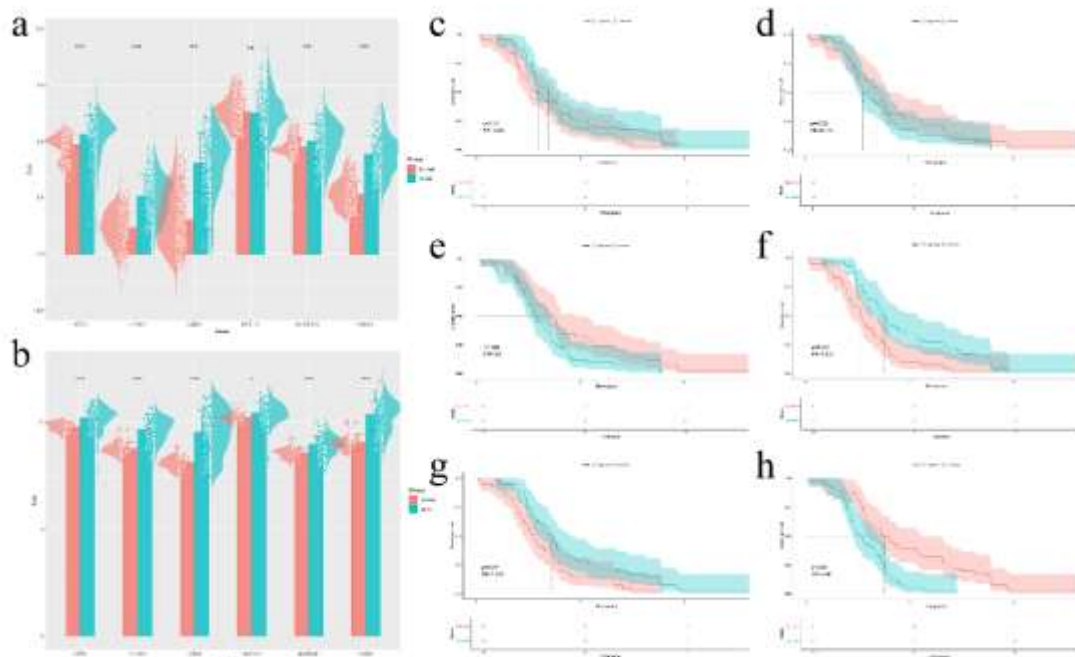


Figure9 Validation of the above analysis results using the GEO dataset. a. Analysis of the expression levels of each candidate gene using TCGA glioblastoma data and GTEx normal sample data; b. Validation of the expression levels of each candidate gene using the GEO validation dataset; c-h. Further analysis of the prognostic impact of each gene on glioblastoma patients in the GEO dataset and construction of Kaplan-Meier survival curves

Validation of Marker Genes Expression Levels in Clinical GBM Samples

To further validate the practical applicability of these marker genes, we collected clinical GBM samples and adjacent tumor tissues for fluorescent quantitative PCR validation. Using GraphPad and performing T-test analysis, the results showed that DCTD ($p=0.0123$), SEC11C ($p<0.0001$),

OSMR ($p=0.0465$), SMARCB1 ($p=0.0392$), TWSG1 ($p=0.0019$), and IL17RD ($p=0.0035$) were all relatively highly expressed in GBM tissues (**Figure10a-f**), which was consistent with the previous bioinformatics analysis results, indicating that these marker genes hold significant potential implications for the diagnosis and clinical treatment of GBM.

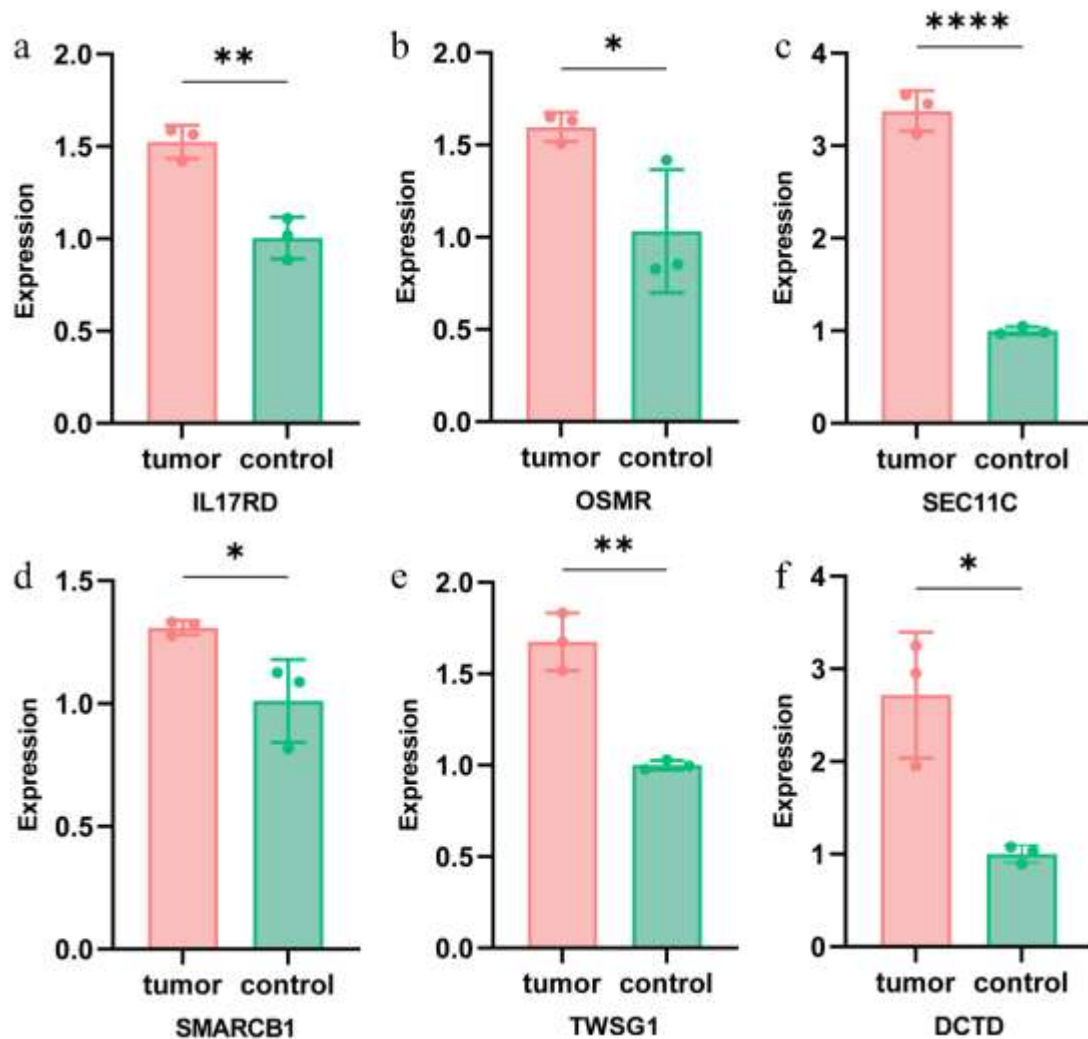


Figure10 Quantitative PCR validation of the expression levels of each candidate gene using clinical glioblastoma samples. a-f. IL17RD, OSMR, SEC11C, SMARCB1, TWSG1, and DCTD respectively

Materials and Methods

Data Acquisition and Preprocessing

All original analysis data for this study were obtained from the TCGA, UCSC Xena, and GEO databases. We downloaded data from 176 GBM samples (including 172 tumor samples and 4 normal samples) along with clinical phenotype data from TCGA (<https://portal.gdc.cancer.gov/>) and extracted RNA-seq expression data in Transcripts Per Million (TPM) format. We excluded samples from the clinical phenotype data that lacked survival status or survival time and retained sample names, survival status, and survival time, resulting in 160 clinical samples. We then downloaded the normal brain tissue

expression data (in TPM format) and phenotype data from the GTEx database (<https://xenabrowser.net/datapages/>). The GBM data from TCGA and the normal brain tissue data from GTEx (500 normal samples) were merged, batch effects were removed, and the data were normalized and log2-transformed to create the test set. Furthermore, we downloaded the GSE43378, GSE4290, and GSE83300 datasets from the GEO database (<https://www.ncbi.nlm.nih.gov/geo/>) for validation. Additionally, 16 disulfide cell death-related genes (H. Liu & Tang, 2023) were obtained from the literature for subsequent analysis.

Differential Analysis and Selection of Independent Prognostic Genes

We performed differential analysis using the merged TCGA GBM data and GTEx normal data in R 4.4.0 with the limma package to identify differentially expressed genes (DEGs). In the analysis, the thresholds for selection were set at an adjusted p-value < 0.05 and $|\log_2(\text{FC})| > 1$. We then intersected the obtained DEGs with the 16 disulfide death genes to identify differentially expressed disulfide death genes. These genes were further analyzed through Pearson correlation analysis with all genes, selecting genes with correlation coefficients $R > 0.5$ and p-values < 0.05 . The Lasso analysis of these correlated genes was performed using the glmnet package in R, and the optimal cross-validation result was obtained at $\lambda = -2.681073$. Subsequently, a univariate analysis model was constructed using the patients' survival status and survival time to further screen for independent prognostic genes. Genes with a p-value < 0.05 from the univariate analysis were included in the subsequent analysis.

Multiple Machine Learning Algorithms Analysis

We applied four machine learning algorithms, SVM-RFE, Lasso, Random Forest, and XGBoost, to further filter the independent prognostic genes. Genes identified by all four algorithms were selected for further analysis

ROC Curve

ROC curves were constructed to evaluate the predictive performance of the Support Vector

Machine (SVM) algorithm, and multi-gene ROC curves were generated to assess the diagnostic sensitivity and specificity of the selected genes.

Survival Curve Analysis

We used TCGA tumor data and its clinical information as the test set to assess the prognostic impact of biomarker genes in GBM. Additionally, the GSE43378 and GSE83300 datasets were used as validation sets to further analyze the prognostic effect of these biomarkers in GBM. In R 4.4.0, the median expression value of each biomarker gene was used as the threshold to divide the samples into high and low expression groups, followed by survival analysis using the survival package.

Single-gene Gene Set Enrichment Analysis (GSEA)

We performed this analysis in R 4.4.0 using the GSEA package to further explore the relevant signaling pathways of these six biomarker genes. TCGA tumor expression data were used for this analysis. First, we calculated the correlation of each biomarker gene with all other genes and ranked them according to the log fold change (logFC) for subsequent analysis. Then, we downloaded the KEGG pathway dataset from the GSEA database (<https://www.gsea-msigdb.org/gsea/index.jsp>) as the standard for enrichment evaluation. The top 5 pathways from the final result were selected for display, and Table S1 lists the GSEA enrichment results for each biomarker gene.

Table S1

Description	pvalue	p.adjust
REFERENCE_CYTOKINE_JAK_STAT_SIGNALING_PATHWAY	7.2224680536 8083e-06	0.00012384363 9662769
REFERENCE_TRANSLATION_INITIATION	1.3036172596 0809e-05	0.00012384363 9662769
REFERENCE_GENE_SILENCING_BY_METHYLATION_OF_H3K27_AND_UBIQUITINATION_OF_H2AK119	0.0004868545 27222385	0.00308341200 574177
REFERENCE_RAB7_REGULATED_MICROTUBULE_MINUS_END	0.0061242092	0.02908999375

_DIRECTED_TRANSPORT	1087074	1636
REFERENCE_RETROGRADE_AXONAL_TRANSPORT	0.0122390088 021194	0.04650823344 80537
DCTD		
Description	pvalue	p.adjust
REFERENCE_MITOCHONDRIAL_COMPLEX_UCP1_IN_THERMOGENESIS	4.884803228 95737e-08	1.172352774 94977e-06
VARIANT_MUTATION_CAUSED_ABERRANT_SNCA_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	8.788771039 08839e-07	1.054652524 69061e-05
REFERENCE_CYTOKINE_JAK_STAT_SIGNALING_PATHWAY	1.630536420 55594e-06	1.304429136 44475e-05
REFERENCE_ELECTRON_TRANSFER_IN_COMPLEX_I	2.992557114 36612e-06	1.436427414 89574e-05
VARIANT_MUTATION_INACTIVATED_PINK1_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	2.992557114 36612e-06	1.436427414 89574e-05
VARIANT_MUTATION_CAUSED_ABERRANT_TDP43_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	7.780720277 04476e-06	3.112288110 8179e-05
VARIANT_MUTATION_CAUSED_ABERRANT_ABETA_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	1.120106804 77625e-05	3.840366187 8043e-05
REFERENCE_WNT_SIGNALING_PATHWAY	0.001351492 44062096	0.004054477 32186287
VARIANT_FZD7_OVEREXPRESSION_TO_WNT_SIGNALING_PATHWAY	0.003092683 22786953	0.008247155 27431873
VARIANT_LRP6_OVEREXPRESSION_TO_WNT_SIGNALING_PATHWAY	0.005193958 26466617	0.012465499 8351988
VARIANT_MUTATION_CAUSED_ABERRANT_SNCA_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.013273585 0125638	0.028960549 118321
IL17RD		
Description	pvalue	p.adjust
REFERENCE_TRANSLATION_INITIATION	6.48843638660036e-10	1.16791854958806e-08
REFERENCE_CYTOKINE_JAK_STAT_SIGNALING_PATHWAY	0.000113152207308 795	0.001018369865779 15
REFERENCE_ITGA_B_RHOG_RAC_SIGNALING_PATHWAY	0.006997912257974 45	0.041987473547846 7
TWSG1		
Description	pvalue	p.adjust
REFERENCE_TRANSLATION_INITIATION	4.8789454158 389e-09	1.4148941705 9328e-07
REFERENCE_CYTOKINE_JAK_STAT_SIGNALING_PATHWAY	1.3893630374 8461e-08	2.0145764043 5268e-07
REFERENCE_ITGA_B_TALIN_VINCULIN_SIGNALING_PATHWAY	3.0437262717 7188e-07	2.9422687293 7948e-06
REFERENCE_GENE_SILENCING_BY_METHYLATION_OF_H3K27_AND_UBIQUITINATION_OF_H2AK119	1.7104919149 3294e-06	1.2401066383 2638e-05
REFERENCE_ITGA_B_RHOG_RAC_SIGNALING_PATHWAY	0.0001446906 37617133	0.0008392056 98179374
REFERENCE_ITGA_B_FAK_CDC42_SIGNALING_PATHWAY	0.0002076317 52230362	0.0010035534 6911342
REFERENCE_ITGA_B_FAK_RAC_SIGNALING_PATHWAY	0.0002936815	0.0012166808

	92462643	8305952
REFERENCE_MITOCHONDRIAL_COMPLEX_UCP1_IN_THERMOGENESIS	0.0006341354	0.0022987408
	12960826	7198299
REFERENCE_ADRB3_UCP1_SIGNALING_PATHWAY	0.0043791979	0.0141107488
	1657874	423093
REFERENCE_TIGHT_JUNCTION_ACTIN_SIGNALING_PATHWAY	0.0132825059	0.0385192672
Y	321365	031958
OSMR		
Description	pvalue	p.adjust
REFERENCE_MITOCHONDRIAL_COMPLEX_UCP1_IN_THERMOGENESIS	4.797572220	1.1993930551
	56245e-25	4061e-23
REFERENCE_ELECTRON_TRANSFER_IN_COMPLEX_I	6.697250306	5.5810419220
	497e-16	8084e-15
VARIANT_MUTATION_CAUSED_ABERRANT_SNCA_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	6.697250306	5.5810419220
	497e-16	8084e-15
VARIANT_MUTATION_INACTIVATED_PINK1_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	1.712510847	1.0703192795
	33133e-15	8208e-14
VARIANT_MUTATION_CAUSED_ABERRANT_TDP43_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	3.379919783	1.6899598918
	66445e-15	3222e-14
VARIANT_MUTATION_CAUSED_ABERRANT_ABETA_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	2.498603468	1.0410847785
	58677e-14	7782e-13
VARIANT_MUTATION_CAUSED_ABERRANT_SOD1_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	9.709746300	0.0003467766
	21194e-05	53578998
REFERENCE_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.000298097	0.0009315545
	441343685	04199015
REFERENCE_WNT_SIGNALING_PATHWAY	0.000386523	0.0010736759
	327648775	1013549
VARIANT_MUTATION_INACTIVATED_UBQLN2_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.000495843	0.0012396080
	238216389	9554097
VARIANT_MUTATION_INACTIVATED_VCP_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.000917254	0.0020846686
	212366556	6446945
VARIANT_MUTATION_CAUSED_ABERRANT_ABETA_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.001507767	0.0031411821
	42193238	2902578
REFERENCE_TRANSLATION_INITIATION	0.001789140	0.0034406542
	18888924	0940239
VARIANT_MUTATION_CAUSED_ABERRANT_HTT_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.002827044	0.0050482943
	8508062	7643965
REFERENCE_ADRB3_UCP1_SIGNALING_PATHWAY	0.006577272	0.0109621212
	72334211	055702
REFERENCE_TRANSCRIPTION_COUPLED_NER	0.016199794	0.0253121789
	5196434	369428
REFERENCE_RAB7_REGULATED_MICROTUBULE_MINUS_END_DIRECTED_TRANSPORT	0.032035432	0.0471109299
	3814908	727806
SEC11C		
Description	pvalue	p.adjust
REFERENCE_TRANSLATION_INITIATION	5.169956828	1.550987048
	81154e-19	64346e-17
REFERENCE_PRE_IC_FORMATION	1.947246677	2.920870016
	6046e-07	4069e-06
REFERENCE_CYTOKINE_JAK_STAT_SIGNALING_PATHWAY	5.434043606	5.434043606
	93702e-06	93702e-05

REFERENCE_ORIGIN_UNWINDING_AND_ELONGATION	1.888567041 30004e-05	0.000113314 022478002
REFERENCE_MITOCHONDRIAL_COMPLEX_UCP1_IN_THERMOGENESIS	1.707329328 34068e-05	0.000113314 022478002
VARIANT_MUTATION_CAUSED_ABERRANT_HTT_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	4.583692274 73168e-05	0.000196443 954631358
VARIANT_MUTATION_CAUSED_ABERRANT_SNCA_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	4.583692274 73168e-05	0.000196443 954631358
REFERENCE_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	8.562517092 25857e-05	0.000256875 512767757
VARIANT_MUTATION_INACTIVATED_UBQLN2_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	8.562517092 25857e-05	0.000256875 512767757
VARIANT_MUTATION_INACTIVATED_VCP_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	8.562517092 25857e-05	0.000256875 512767757
VARIANT_MUTATION_CAUSED_ABERRANT_ABETA_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.000467535 029320387	0.001275095 53451015
VARIANT_MUTATION_CAUSED_ABERRANT_SOD1_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.000726063 305346474	0.001815158 26336619
REFERENCE_ITGA_B_TALIN_VINCULIN_SIGNALING_PATHWAY	0.001104714 76371169	0.002549341 76241159
VARIANT_SCRAPIE_CONFORMATION_PRPSC_TO_26S_PROTEASOME_MEDIATED_PROTEIN_DEGRADATION	0.001593370 17005588	0.003414364 65011974
REFERENCE_GENE_SILENCING_BY_METHYLATION_OF_H3K27_AND_UBIQUITINATION_OF_H2AK119	0.002404204 31727433	0.004808408 63454865
REFERENCE_ELECTRON_TRANSFER_IN_COMPLEX_I	0.019519979 13022	0.032533298 5503666
VARIANT_MUTATION_CAUSED_ABERRANT_SNCA_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	0.019519979 13022	0.032533298 5503666
VARIANT_MUTATION_CAUSED_ABERRANT_TDP43_TO_ELECTRON_TRANSFER_IN_COMPLEX_I	0.019519979 13022	0.032533298 5503666

SMARCB1

Single-gene Gene Set Variation Analysis (GSVA)

We conducted this analysis in R 4.4.0 using the GSVA package, a method for analyzing gene set variation. First, we downloaded the classic 50 KEGG cancer signaling pathway datasets from the GSEA database (<https://www.gsea-msigdb.org/gsea/index.jsp>) as the standard for enrichment analysis. Then, in the GBM data, we divided the samples into high and low expression groups based on the median expression value of each biomarker gene and performed differential analysis using the limma package. Pathways with t-values greater than 0 and p-values less than 0.05 were categorized as upregulated, while pathways

with t-values less than 0 and p-values less than 0.05 were categorized as downregulated.

Differential Analysis of GSVA-Enriched Pathways and Correlation with Marker Genes

To further explore potential signaling pathways regulating biomarker genes, we used the limma package in R 4.4.0 to analyze differential signaling pathways between GBM and normal samples in the merged TCGA and GTEx data. We also analyzed the correlation between these differential signaling pathways and the biomarker genes. First, we set the threshold for significant differential signaling pathways as $|\log FC| = 2$ and $p < 0.05$. We then analyzed the correlation between each biomarker gene and these significant differential signaling pathways, as well

as the correlation between the pathways themselves.

Drug Prediction and ceRNA Network Construction

We used the DSigDB database (<https://dsigdb.tanlab.org/DSigDBv1.0/>) to predict the target drugs for these biomarker genes. Additionally, we constructed the mRNA-miRNA interaction network for these biomarker genes using the miRDB database (<https://mirdb.org/>) and the TargetScan database (https://www.targetscan.org/vert_80/). We also built the miRNA-lncRNA interaction network using the starBase database (<https://rnasysu.com/encori/index.php>). These target drug networks and ceRNA networks for the biomarker genes were visualized using Cytoscape software.

Marker Genes Expression Analysis

We used the merged TCGA and GTEx data as the test set, and the GSE4290 dataset as the validation set, to analyze the differential expression of each biomarker gene between GBM and normal samples.

Clinical Samples Acquisition

The clinical samples used to validate biomarker gene expression were obtained from two GBM samples and two adjacent tumor tissues surgically resected by the Neurosurgery Department at Qingdao University Affiliated Hospital. This study was approved by the Ethics Committee of Qingdao University Affiliated Hospital.

Fluorogenic Quantitative PCR Analysis

Total RNA from clinical samples was extracted using the TRNzol Total RNA Extraction Reagent Kit (TianGen Biotech, Beijing). The forward and reverse primers for these biomarker genes are listed as follows: “h-DCTD-F TTGCCTTG GAGAAGGACAGCAG”, “h-DC D-R ACT ACAGCCTTTCACATCGGTCG”, “h-SEC11C-F GGCTGGAAAA GAAGGACGT GGT”, “h-SEC 11C-R GCACCCATTACAG CCAAAGAGC”, “h-OSMR-F CAGGTGTTCC TACCAAATCT GCG”, “h-OSMR-RAATCCACCCTCTGT GC CTGCAA”, “h-SMARCB1-F GGCATCAGAAG ACCTACGCCTT”, “h-SMARCB1-R CTCCATC TCAGCGTCTGTC AGA”, “h-TWSG1-F CTT TGGGACGAGTGCTGTGACT”, “h-TWSG1-R GAGAAGGGATCGGTTTCATGCAG”, “h-IL1 7RD-F GAATGTCGTCCAGTGTTTCGCC”, “h-IL17RD-R GATGAACTGGGACTCGTGGAT C”, “h-GAPDH-F GACCTGACCTGCCGTCTA”, and “h-GAPDH-R AGGAGTGGGTGTCGCTGT”, which is listed in Table S2. mRNA reverse transcription was performed using the cDNA Synthesis Kit (TaKaRa, RR047A), and the resulting cDNA was stored at -20°C for later use. Quantitative PCR was then performed using the Quantitative PCR Kit (TaKaRa, RR820A). The cDNA samples were diluted 10-fold and used as templates for PCR detection (ABI 7500 Fluorescent Quantitative PCR Instrument). The data were saved, and GAPDH was used as the internal control for relative quantification. Statistical analysis was performed using the T-test in GraphPad.

Table S2

Gene Name	Primer Sequence	Length
h-DCTD-F	TTGCCTTG GAGAAGGACAGCAG	132
h-DCTD-R	ACTACAGCCTTTCACATCGGTCG	
h-SEC11C-F	GGCTGGAAAA GAAGGACGTGGT	157
h-SEC11C-R	GCACCCATTACAGCCAAAGAGC	
h-OSMR-F	CAGGTGTTCTTACCAAATCTGCG	144

h-OSMR-R	AATCCACCCTCTGTGCCTGCAA	
h-SMARCB1-F	GGCATCAGAAGACCTACGCCTT	
h-SMARCB1-R	CTCCATCTCAGCGTCTGTCAGA	116
h-TWSG1-F	CTTTGGGACGAGTGCTGTGACT	
h-TWSG1-R	GAGAAGGGATCGGTTCATGCAG	163
h-IL17RD-F	GAATGTCGTCCAGTGTTTCGCC	
h-IL17RD-R	GATGAACTGGGACTCGTGGATC	119
h-GAPDH-F	GACCTGACCTGCCGTCTA	
h-GAPDH-R	AGGAGTGGGTGTCGCTGT	83

Discussion

GBM is one of the most aggressive intracranial tumors, and the current standard treatment remains surgery combined with radiotherapy and chemotherapy (Barbosa et al., 2014; Yi et al., 2019). However, due to the unique location of GBM in the human body and the blood-brain barrier's blocking effect on most small-molecule drugs, the median survival rate for GBM patients is only around 14 months, with a 5-year survival rate of 4-5% (B. Huang et al., 2021; Lu et al., 2022). Therefore, it is crucial to find more effective treatments for GBM and explore diagnostic biomarkers for this disease. Increasing evidence shows that there are multiple biomarkers in GBM that play an important role in its treatment. One study revealed that TRIM7 is highly expressed in GBM and promotes tumor progression. However, silencing TRIM7 can increase GBM sensitivity to temozolomide by regulating NCOA4-mediated ferroptosis (Li et al., 2022). Another study showed that LGALS3 promotes GBM resistance to ionizing radiation and temozolomide, and it is an independent prognostic factor for GBM patients. It may be a marker gene to improve therapeutic resistance and increase patient survival (Wang et al., 2019). Therefore, further identification of marker genes related to the pathogenesis of GBM and investigation of their prognostic value may

provide novel targeted therapeutic directions for clinical treatment.

Using bioinformatics methods to screen disease-specific genes and explore molecular mechanisms related to the disease is an effective research approach. This method not only allows for the observation of general regulatory mechanisms in disease progression but also provides clear research directions for subsequent molecular experiments. In this study, we combined bioinformatics and machine learning to investigate the mechanisms of disulfidptosis - related genes in GBM. Our results suggest that six marker genes, including DCTD, IL17RD, SMARCB1, TWSG1, OSMR, and SEC11C, are of significant potential prognostic value for GBM patients. Through Gene Set Enrichment Analysis (GSEA) and Gene Set Variation Analysis (GSVA), we found that these genes are closely associated with multiple tumor-related signaling pathways, including the IL6_JAK_STAT3, P53, PI3K_AKT, and KRAS signaling pathways. Interestingly, except for SEC11C, the genes DCTD, TWSG1, OSMR, SMARCB1, and IL17RD are significantly enriched in the JAK_STAT3 signaling pathway, and differential analysis showed that the JAK_STAT3 signaling pathway is also significantly different between GBM and normal tissues. As we all know, the STAT protein family includes STAT1, STAT2,

STAT3, STAT4, STAT5A, STAT5B, and STAT6 (S. J. Liu, Dang, Lim, Feng, & Maher, 2021). Among these, STAT3 has the most characteristic oncogenic activity (H. Liu & Tang, 2023). Abnormal expression of STAT3 can promote angiogenesis, immune evasion, and apoptosis resistance in GBM by regulating the transcription of related genes. These signaling targets mainly include Bcl-xL, Bcl-2, cyclin D1, and c-Myc (H. Huang, Weng, & Chen, 2020). Moreover, numerous studies have shown that the STAT3 signaling pathway is involved in regulating GBM drug resistance. For example, activation of the STAT3 pathway can promote TMZ resistance in tumor cells by inhibiting the degradation of MGMT in GBM cells (Jia et al., 2016). Additionally, studies have shown that PIKE-A can activate the phosphorylation level of STAT3, suppressing the transcriptional activity of 6-phosphogluconate dehydrogenase and leading to tumor cell metabolic reprogramming, ultimately contributing to TMZ resistance (He et al., 2020). This provides direction for our subsequent research.

We then validated the expression of these marker genes in GBM using the GSE43378, GSE4290, and GSE83300 datasets and performed survival analysis. Finally, we validated the expression of these genes through quantitative real-time PCR, and the results were consistent with the trends observed in the TCGA dataset and validation datasets. These findings suggest that these six marker genes may potentially regulate the progression and drug resistance of GBM through the JAK-STAT3 signaling pathway and have reliable clinical research value.

However, there are some limitations in this study. First, the relatively small sample size of the datasets used may affect the generalizability of the results. Second, this study mainly utilized bioinformatics methods, and only quantitative real-time PCR was used for experimental validation. There is a lack of further validation

through various *in vitro* and *in vivo* experiments, which reduces the reliability of the study. Moreover, this study only explored biomarkers for GBM and did not further investigate the specific signaling pathways and biological processes these biomarkers are involved in. These are the limitations of this study. To address these shortcomings, future research should focus on the following aspects: First, conduct various *in vitro* experiments to validate the specific biological functions of these marker genes, delving into the specific signaling pathways and drug resistance mechanisms in GBM. Second, validate the roles of these marker genes in animal models to provide new potential targeted therapeutic directions for the clinical treatment of GBM.

Conclusion

In conclusion, this study provides new insights into the molecular mechanisms of potential diagnostic biomarkers for GBM, including the diagnostic and prognostic efficacy of these marker genes in GBM, especially the potential value of signaling pathways (such as the JAK-STAT3 pathway) that these marker genes may be involved in, in the development, progression, and drug resistance of GBM. The ceRNA network of these marker genes and the prediction network for targeted drugs offer new directions for targeted therapy in GBM, providing fresh perspectives for clinical treatment. For future research, we should prioritize the validation of the specific biological functions of these marker genes, elucidating the specific signaling pathways they are involved in and evaluating the efficacy of targeted therapies, and providing potential therapeutic targets for clinical treatment.

Statements and Declarations

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Competing Interests

The authors have no relevant financial or non-financial interests to disclose.

Author Contributions

All authors contributed to the study. Conceptualization, Visualization and Writing-original draft were performed by Daikang Xu; Formal Analysis and Methodology were performed by Guangyu Du; Data Curation and Software were achieved by Jingxian Sun; Resources and Validation were finished by Xiaolei Lan and Zhiyong Yan; Project Administration and Supervision were achieved by Jianpeng Wang.

Data Availability

The datasets analysed during the current study are available from the TCGA database (<https://portal.gdc.cancer.gov/>), GEO database (<https://www.ncbi.nlm.nih.gov/geo/>), and GTEx database (<https://portal.gdc.cancer.gov/>). Data from the analysis can be obtained from the corresponding author upon reasonable request.

Ethics Declarations

This study was performed in line with the principles of the Declaration of Helsinki. Approval was granted by the Ethics Committee of Affiliated hospital of Qingdao University (Date May 8th, 2025; Approval number QYFYWZ LL30036).

References

1. Barbosa, B. J. A. P., Mariano, E. D., Batista, C. M., Marie, S. K. N., Teixeira, M. J., Pereira, C. U., . . . Lepski, G. A. (2014). Intraoperative assistive technologies and extent of resection in glioma surgery: a systematic review of prospective controlled studies. *Neurosurgical Review*, 38(2), 217-227. doi:10.1007/s10143-014-0592-0
2. Goji, T., Takahara, K., Negishi, M., & Katoh, H. (2017). Cystine uptake through the cystine/glutamate antiporter xCT triggers glioblastoma cell death under glucose deprivation. *Journal of Biological Chemistry*, 292(48), 19721-19732. doi:10.1074/jbc.M117.814392
3. He, X., Sheng, J., Yu, W., Wang, K., Zhu, S., & Liu, Q. (2020). LncRNA MIR155HG Promotes Temozolomide Resistance by Activating the Wnt/ β -Catenin Pathway Via Binding to PTBP1 in Glioma. *Cellular and Molecular Neurobiology*, 41(6), 1271-1284. doi:10.1007/s10571-020-00898-z
4. Huang, B., Li, X., Li, Y., Zhang, J., Zong, Z., & Zhang, H. (2021). Current Immunotherapies for Glioblastoma Multiforme. *Frontiers in Immunology*, 11. doi:10.3389/fimmu.2020.603911
5. Huang, H., Weng, H., & Chen, J. (2020). m6A Modification in Coding and Non-coding RNAs: Roles and Therapeutic Implications in Cancer. *Cancer Cell*, 37(3), 270-288. doi:10.1016/j.ccell.2020.02.004
6. Jia, X., Wang, Z., Qiu, L., Yang, Y., Wang, Y., Chen, Z., . . . Yu, L. (2016). Upregulation of LncRNA-HIT promotes migration and invasion of non-small cell lung cancer cells by association with ZEB1. *Cancer Medicine*, 5(12), 3555-3563. doi:10.1002/cam4.948
7. Joly, J. H., Delfarah, A., Phung, P. S., Parrish, S., & Graham, N. A. (2020). A synthetic lethal drug combination mimics glucose deprivation-induced cancer cell death in the

- presence of glucose. *Journal of Biological Chemistry*, 295(5),1350-1365. doi:10.1016/s0021-9258(17)49891-7
8. Lee, S., Weiss, T., Bühler, M., Mena, J., Lottenbach, Z., Wegmann, R., . . . Snijder, B. (2024). High-throughput identification of repurposable neuroactive drugs with potent anti-glioblastoma activity. *Nature Medicine*, 30(11), 3196-3208. doi:10.1038/s41591-024-03224-y
 9. Li, K., Chen, B., Xu, A., Shen, J., Li, K., Hao, K., . . . Wang, Z. (2022). TRIM7 modulates NCOA4-mediated ferritinophagy and ferroptosis in glioblastoma cells. *Redox Biology*, 56. doi:10.1016/j.redox.2022.102451
 10. Liu, H., & Tang, T. (2023). Pan-cancer genetic analysis of disulfidptosis-related gene set. *Cancer Genet*, 278-279,91-103. doi:10.1016/j.cancergen.2023.10.001
 11. Liu, S. J., Dang, H. X., Lim, D. A., Feng, F. Y., & Maher, C. A. (2021). Long noncoding RNAs in cancer metastasis. *Nature Reviews Cancer*, 21(7), 446-460. doi:10.1038/s41568-021-00353-1
 12. Liu, X., Nie, L., Zhang, Y., Yan, Y., Wang, C., Colic, M., . . . Gan, B. (2023). Actin cytoskeleton vulnerability to disulfide stress mediates disulfidptosis. *Nature Cell Biology*, 25(3), 404-414. doi:10.1038/s41556-023-01091-2
 13. Lu, G., Wang, X., Li, F., Wang, S., Zhao, J., Wang, J., . . . Wei, W. (2022). Engineered biomimetic nanoparticles achieve targeted delivery and efficient metabolism-based synergistic therapy against glioblastoma. *Nature Communications*, 13(1). doi:10.1038/s41467-022-31799-y
 14. Ma, Q., Jiang, H., Ma, L., Zhao, G., Xu, Q., Guo, D., . . . Lu, Z. (2023). The moonlighting function of glycolytic enzyme enolase-1 promotes choline phospholipid metabolism and tumor cell proliferation. *120(15)*, e2209435120. doi:doi:10.1073/pnas.2209435120
 15. Mandal, M., Banerjee, I., & Mandal, M. (2025). Effective approaches in conquering chemoresistance of glioblastoma: potential for nanoformulations. *Drug Delivery and Translational Research*. doi:10.1007/s13346-025-01859-z
 16. Qin, A., Musket, A., Hilton, B., Preiszner, J., Krenciute, G., Berens, M. E., . . . Xie, Q. (2025). Efficacy of MET-targeting CAR T cells against glioblastoma patient-derived xenograft models. *Journal of Translational Medicine*, 23(1). doi:10.1186/s12967-025-06475-6
 17. Rodríguez-Camacho, A., Flores-Vázquez, J. G., Moscardini-Martelli, J., Torres-Ríos, J. A., Olmos-Guzmán, A., Ortiz-Arce, C. S., . . . Moreno-Jiménez, S. (2022). Glioblastoma Treatment: State-of-the-Art and Future Perspectives. *International Journal of Molecular Sciences*, 23(13). doi:10.3390/ijms23137207
 18. Tamai, S., Ichinose, T., Tsutsui, T., Tanaka, S., Garaeva, F., Sabit, H., & Nakada, M. (2022). Tumor Microenvironment in Glioma Invasion. *Brain Sciences*, 12(4). doi:10.3390/brainsci12040505
 19. Tummers, B., & Green, D. R. (2017). Caspase-8: regulating life and death. *Immunological Reviews*, 277(1),76-89. doi: 10.1111/imr.12541
 20. Wang, H., Song, X., Huang, Q., Xu, T., Yun, D., Wang, Y., . . . Chen, J. (2019). LGALS3 Promotes Treatment Resistance in Glioblastoma and Is Associated with Tumor Risk and Prognosis. *Cancer Epidemiology, Biomarkers & Prevention*, 28(4), 760-769. doi:10.1158/1055-9965.Epi-18-0638
 21. Xu, J., Guo, Y., Ning, W., Wang, J., Chen, Y., Liu, D., . . . Zhang, H. (2025). Comprehensive analysis of heat shock proteins in glioma revealed the association with glioma-associated myeloid cells. *Genes & Immunity*. doi:10.1038/s41435-025-00327-5
 22. Yi, H. G., Jeong, Y. H., Kim, Y., Choi, Y. J., Moon, H. E., Park, S. H., . . . Cho, D. W.

- (2019). A bioprinted human-glioblastoma-on-a-chip for the identification of patient-specific responses to chemoradiotherapy. *Nat Biomed Eng*, 3(7), 509-519. doi:10.1038/s41551-019-0363-x
23. Zhang, S., Wang, Y., Sun, B., Zhu, S., Jia, Z., Liu, L., & Liu, L. (2025). Regulation of Glycolysis by SMAD5 in Glioma Cells: Implications for Tumor Growth and Apoptosis. *Neurochemical Research*, 50(2). doi:10.1007/s11064-025-04352-8
24. Zhao, G., Zhao, Z., Xia, M., Xiao, L., Zhu, B., Wang, H., . . . Di, J. (2023). CPEB2 inhibit cell proliferation through upregulating p21 mRNA stability in glioma. *Scientific Reports*, 13(1). doi:10.1038/s41598-023-50848-0