Research Article

PCMT1 Is a Potential Prognostic Biomarker and Is Correlated with Immune Infiltrates in Breast Cancer

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Background. Protein-L-isoaspartate (D-aspartate) O-methyltransferase (PCMT1) is involved in the occurrence and development of a variety of malignant tumors. However, the prognostic value of PCMT1 in breast cancer remains unclear.

Methods. Based on the Cancer Genome Atlas database, we assessed the correlation between the expression of PCMT1 and prognosis, immune invasion, and tumor mutation burden in a variety of cancers. The expression level, mutation, immune correlation, and coexpression of PCMT1 in breast cancer were studied using the following databases: UALCAN database, Human Protein Atlas database, cBioPortal database, TIMER database, and LinkedOmics database. Kaplan–Meier Plotter was used for survival analysis. Receiver operating characteristic (ROC) curves and nomograms were drawn using the R software package. 

Results. Pancancer analysis showed that PCMT1 is highly expressed in a variety of cancers and is significantly related to the prognosis of a variety of cancers. PCMT1 is significantly related to the tumor mutation burden of a variety of cancers. PCMT1 is significantly high in breast cancer, and it is significantly related to the abundance of immune infiltration. Survival analysis revealed that high PCMT1 expression is significantly associated with shorter overall survival (OS), relapse-free survival (RFS), and postprogression survival (PPS) in breast cancer patients. ROC curves and nomograms verify the effectiveness of PCMT1 as a prognostic biomarker for breast cancer.

Conclusions. PCMT1 can be used as a potential prognostic biomarker of breast cancer, and it is significantly related to the abundance of breast cancer immune infiltration.

1. Introduction

Breast cancer (BRCA) is one of the most common malignant tumors in women worldwide [1]. The annual incidence and fatality rate of breast cancer ranks at the forefront of all cancers [2]. Statistics show that among the most common cancers diagnosed in American women in 2021, breast cancer alone accounts for 30% of female cancers. In 2021, there will be 284,200 new breast cancer patients in the USA, of which 44,130 deaths are estimated [3]. However, there are still many unclear parts of the specific molecular biological mechanism of breast cancer occurrence and development. In addition, the early diagnosis rate of breast cancer is low, and the side effects of treatment are large, resulting in the overall effect of clinical breast cancer treatment being unsatisfactory [4–6].

At present, the main means to reduce the incidence and mortality of breast cancer is to find specific diagnostic criteria and increase the sensitivity and specificity of early diagnosis [7, 8]. In recent years, biomarkers have attracted wide attention because of their ability to predict tumor development, prognosis, and response to treatment [9]. The emergence of molecular biomarkers of breast cancer is aimed at improving the early diagnosis rate of breast cancer, predicting the effects of related prognostic treatments, and providing better clinical guidance [10]. With the update of detection technology and the continuous development of immunotherapy, an increasing number of breast cancer biomarkers have been used clinically. However, based on the limited clinical specificity and sensitivity, the development of new breast cancer biomarkers is still very urgent [11].

Protein-L-isoaspartate (D-aspartate) O-methyltransferase (PCMT1) is an S-adenosylmethionine-dependent methyltransferase. The main functions of PCMT1 are to initiate the repair of damaged proteins by catalyzing
Figure 1: (a) Gene expression: the expression distribution of PCMT1 gene in tumor tissues and normal tissues. (b) Differential expression of PCMT1 in pancancer (TIMER). (c) Immune correlation: a heat map of the correlation between PCMT1 and multiple tumor tissues. (d) Forest plot: the correlation between PCMT1 gene and prognosis in multiple tumors.
Expression of PCMT1 in BRCA based on individual cancer stages

(a) mRNA expression of PCMT1 in BRCA

(b) Promoter methylation level of PCMT1 in BRCA

(c) Protein expression of PCMT1 in Breast cancer

Figure 2: Continued.
methyl esterification of L-isoaspartyl and D-aspartyl residues produced by spontaneous isomerization and racemization of L-aspartyl and L-asparaginyl residues in aging peptides and proteins. Previous studies have shown that strong PCMT1 expression is not only a predictive marker for poor prognosis of surgically removed lung adenocarcinoma but also an unfavorable prognostic biomarker for bladder cancer [12–14]. It participates in cell migration and invasion by regulating epithelial-mesenchymal transition-related genes and acts as an oncogene in bladder cancer [15]. In addition, the latest research shows that PCMT1 promotes the migration and invasion of human U-87 MG and U-251 MG glioblastoma cell lines and plays a key role in the growth of glioblastoma [16]. These studies show that PCMT1 is very important for the occurrence and development of a variety of malignant tumors. Through pancancer and related bioinformatics analysis of PCMT1 in breast cancer, it will help us understand the related functions of PCMT1 in breast cancer and provide new insights for breast cancer drug development and clinical exploration of molecular markers for diagnosis and prognosis.

2. Materials and Method

2.1. Pancancer Analysis of PCMT1. We downloaded a unified standardized pancancer data set from the UCSC (https://xenabrowser.net/) database: TCGA pancancer (PANCAN, N = 10535, G = 60499), and then, we extracted the PCMT1 gene expression data in each sample (sample sources include solid normal tissue and primary tumor tissue). The log2 (x + 0.001) transformation was performed on each expression value. Finally, we also eliminated the cancer species with less than 3 samples and finally obtained the expression data of 26 cancer species [16, 17]. First, we used R software v4.0.3 to compare the expression of PCMT1 in a variety of cancer tissues and normal tissues. The DiffExp module in the TIMER database allows users to study the differential expression of PCMT1 between tumors and adjacent normal tissues in all TCGA tumors. The PCMT1 pancancer prognosis analysis adopts univariate Cox regression analysis and is realized by the “forestplot” R package, using forest plots to display the P value, hazard ratio (HR), and 95% confidence interval (CI) of each variable. To perform a reliable immune correlation assessment, we used the R software package ImmuneDecov, which integrates the six latest algorithms, including TIMER, xCell, MCP-counter,
Figure 3: Continued.
CIBERSORT, EPIC, and quanTiseq [18–20]. SIGLEC15, IDO1, CD274, HAVCR2, PDCD1, CTLA4, LAG3, and PDCD1LG2 are transcripts related to immune checkpoints [21]. We extracted the expression values of these 8 genes and observed the expression of immune checkpoint-related genes. Tumor mutation burden was derived from the article, the immune landscape of cancer, published by Vesteinn Thorsson et al. in 2018 [22]. The Spearman correlation

\[
\text{Log}_2(\text{PCM1 expression}) = 18.72, r = -0.426, p = 0.003, \text{CI}_{95\%} = [0.02, 0.05], p_{\text{corr}} = 0.003
\]

TMB score

\[
\text{Expression} \text{ Wilcox tests } p = 5.1 \times 10^{-5}
\]

True 191 153 395

False 358 395

TIDE score

\[
\text{Expression} \text{ Wilcox tests } p = 5.1 \times 10^{-5}
\]

(f)

Figure 3: (a) Analysis of pancancer mutations of PCMT1. (b) Analysis of PCMT1 mutations in breast cancer. (c) Correlation between PCMT1 and the abundance of immune infiltration in breast cancer. (d) Correlation between PCMT1 and TMB. (e) Correlation between PCMT1 and MSI. (f) The relationship between PCMT1 expression level and ICB response.
**Figure 4: Continued.**

(a) Immune activity scores

(b) Group

(c) Gene expression heatmap
analysis between tumor mutation burden and PCMT1 gene expression was calculated using R software v 4.0.3. The rank-sum test was used to detect the two sets of data, and a P value of <0.05 was considered statistically significant.

2.2. Analysis of PCMT1 Expression in Breast Cancer. We downloaded the unified normalized TCGA-BRCA data set (tumor samples =1092; normal samples =292) from the UCSC (https://xenabrowser.net/) database. The expression data of the PCMT1 gene in each sample were extracted, and log2 (x + 0.001) transformation was performed on each expression value. Expression differences between normal and tumor samples were calculated using R software (version 3.6.4), and significant differences were analyzed using unpaired Wilcoxon rank sum and signed rank tests. And finally, the expression value of PCMT1 in normal breast and breast tumors was obtained. In addition, we downloaded the GSE3744 data set (tumor samples =40; normal samples =7) from the GEO database for validation of differential expression. The UALCAN database (http://ualcan.path.uab.edu/) can be used to analyze cancer omics data [23]. We used TCGA sample analysis to evaluate the expression of PCMT1 protein in breast cancer, the level of PCMT1 methylation, and the correlation with tumor staging. The Human Protein Atlas (HPA) network database (https://www.proteinatlas.org/) further evaluated the protein expression of PCMT1 in clinical breast cancer samples, which contains more than 10 million immunohistochemistry images of various human cells, tissues, and organs [24]. The Cancer Cell Line Encyclopedia (CCLE) database is an online database that can freely explore the genetic information of 947 human tumor cell lines [25]. We used CCLE to evaluate the expression level of PCMT1 in different breast cancer cell lines. The expression level of PCMT1 in breast cancer cell lines was converted from log2 and plotted as a heat map. Breast cancer gene-expression miner (bc-GenExMiner) v4.7 (Updated on June 28, 2021) contains a large amount of breast cancer genome data, which can perform statistical analysis on expression, prognosis, and correlation [26]. The relationship between the PCMT1 gene and the clinicopathologic parameters of breast cancer were analyzed by using bc-GenExMiner v4.7.

2.3. PCMT1 Mutation and Immune Infiltration in Breast Cancer. The cBioPortal for cancer genomics (https://www.cbioportal.org) is based on a multidimensional cancer genomics data set, providing fast, intuitive, and high-quality access to the molecular profile and clinical attributes of large-scale cancer genomics projects [27]. We explored PCMT1 mutations in the clinical breast cancer data set (TCGA, Firehose Legacy, 1108 total samples). TIMER (https://cistrome.shinyapps.io/timer/) aims to systematically analyze the results of immune infiltration of multiple cancers [28]. We used TIMER to evaluate the correlation between the expression of PCMT1 in breast cancer and the abundance of B cells, CD8+ T cells, CD4+ T cells, macrophages, neutrophils, and dendritic cells in breast cancer. The results are displayed as a scatter plot, showing the statistical significance and the purity-corrected part of Spearman’s rho value. We also evaluated the correlation between high and low expression of PCMT1 and breast cancer immune cell infiltration based on ssGSEA (single sample gene set enrichment analysis) and EPIC algorithm. TIP (tracking tumor immunophenotype) can be used to analyze seven anticancer immune states and the proportion of tumor-infiltrating immune cells. The seven anticancer

Figure 4: (a) Correlation between PCMT1 expression and immune cell infiltration in breast cancer (ssGSEA, boxplot). (b) Correlation between PCMT1 expression and seven-step anticancer immune status in breast cancer (TIP database). (c) Correlation between PCMT1 expression and immune cell infiltration in breast cancer (ssGSEA, heatplot). (d) Correlation between PCMT1 expression and immune cell infiltration in breast cancer (EPIC).
immune states include release of cancer cell antigens (step 1), cancer antigen presentation (step 2), priming and activation (step 3), trafficking of immune cells to tumors (step 4), infiltration of immune cells into tumors (step 5), recognition of cancer cells by T cells (step 6), and killing of cancer cells (step 7). We calculated the correlation between high and low expression of PCMT1 and breast cancer immune activity score based on the TIP database [29].

In addition, based on the mRNA sequence data of TCGA breast cancer, we analyzed the Spearman correlation between PCMT1 expression and tumor mutation burden (TMB) and microsatellite instability (MSI). Immune checkpoint blockade (ICB) therapy has completely changed the treatment of human cancer. Based on the expression profile data, we use the tumor immune dysfunction and exclusion (TIDE) algorithm to predict the responsiveness of PCMT1 expression to predictive immune checkpoint inhibitors [30]. A high TIDE score means that immune checkpoint blocking (ICB) has poor efficacy and short survival after ICB treatment.

2.4. PCMT1 Coexpressed Genes in Breast Cancer. The LinkedOmics database (http://www.linkedomics.org/login.php) is a publicly available portal that includes multiomics data from all 32 TCGA cancer types [31]. We used this database to identify genes closely related to PCMT1 and combined the cluego [32] and cluepedia [33] plug-ins in Cytoscape [34] and Metascape [35] to perform functional enrichment analysis of the first 50 coexpressed genes of PCMT1.

2.5. Kaplan–Meier Plotter. Based on the online server Kaplan–Meier (KM) Plotter (https://kmplot.com/analysis/) [36], we evaluated the prognostic value of PCMT1 for overall survival (OS), relapse-free survival (RFS), distant metastasis-free survival (DMFS), and postprogression survival (PPS) in breast cancer. P < 0.05 is considered statistically significant.

2.6. The Influence of the PCMT1 Gene on the Prognosis of Breast Cancer. We downloaded the raw counts and corresponding clinical information of the RNA sequencing data (level 3) of 1,097 breast cancer tumors from the TCGA data set. The log rank was used to test the KM survival analysis to compare the survival differences between the above two or more groups, and timeROC analysis was performed to compare the prediction accuracy and risk score of PCMT1. Gene expression and survival time and survival status in the
TCGA data set were implemented through the R software package ggrisk; the KM survival curve distribution of PCMT1 in the TCGA data set was implemented through the R software packages survival and survminer. The receiver operator characteristic curve (ROC) and area under the curve (AUC) of PCMT1 at different times were determined by the R software package timeROC [37]. For the KM curve, the P value and the HR with 95% CI were obtained by log rank test and univariate Cox proportional hazard regression. All the above analysis methods and R software packages were implemented using v4.0.3 version R software (R Foundation for Statistical Computing, 2020). P < 0.05 was considered statistically significant.

2.7. Construction and Verification of Nomogram. We assessed the impact of PCMT1 and clinical factors such as age, sex, race, PT stage, and PN stage on prognosis. First, univariate and multivariate Cox regression analyses were performed, and forest plots were generated with the “forestplot” R package to display the P value, HR, and 95% CI of each variable. According to the results of multivariate Cox proportional hazard analysis, the R software package “rms” was used to establish a nomogram to predict the total recurrence rate in 3 years. The nomogram provides a graphical representation of these factors, and the prognostic risk of a single patient can be calculated through the points related to each risk factor [38].

![Figure 6: Survival analysis of PCMT1 in breast cancer, including OS, RFS, DMFS, and PPS.](image-url)
3. Results

3.1. Pancancer Analysis Results of PCMT1. We used R software to calculate the expression difference between normal samples and tumor samples in each tumor and used unpaired Wilcoxon rank Sum and signed rank tests to analyze the significance of the difference. We observed that the PCMT1 gene was significantly up-regulated in 10 tumors including colon adenocarcinoma (COAD), tumor: 5.57 ± 0.42; normal: 5.44 ± 0.19, P = 8.4e−4) and breast carcinoma (BRCA, tumor: 6.04 ± 0.60; normal: 5.79 ± 0.20, P = 9.8e−11) (Figure 1(a)). The TIMER database also yielded similar results (Figure 1(b)). Then, we used univariate Cox regression analysis to evaluate the prognosis of PCMT1 for a variety of human cancers. The results are shown in Figure 1(d).

PCMT1 has a good prognostic effect on a variety of cancers, including BRCA (HR: 1.9 (1.3, 2.67), P = 0.001). The results of immune correlation evaluation showed that in addition to adenocarcinoma carcinoma (ACC), lymphoid neoplasm diffuse large B cell lymphoma (DLBC), and uveal melanoma (UVM), PCMT1 has varying degrees of correlation with the immune infiltrating cells of a variety of human cancers (Figure 1(c)). We also evaluated the expression of immune checkpoint-related genes in different tumor tissues (see Supplementary Figure 1). The results showed that with the exception of ACC, cholangiocarcinoma (CHOL), DLBC, and esophageal carcinoma (ESCA), there was a significant correlation between the expression of most tumors and immune checkpoint-related genes. In addition, we also assessed the correlation between tumor mutation burden and PCMT1 expression (See Supplementary Figure 2). The results show that there is a significant positive correlation between PCMT1 expression and tumor mutation burden in multiple tumor types, including ACC, uterine carcinoma (UCS), and BRCA.

3.2. PCMT1 Is Overexpressed in Breast Cancer. Based on the TCGA-BRCA data set, we obtained the differential expression of PCMT1 in breast cancer tissues and normal tissues. The expression of PCMT1 in breast cancer tissues was significantly higher than that in normal tissues (P = 2.2E−49), and this differential expression was also verified in GSE3744 (P = 5.7E−5) (Figure 2(a)). We used the UALCAN online database to compare the expression levels of PCMT1 protein in normal and breast tissues. The protein expression level of PCMT1 in breast cancer tissues (n = 125) was lower than that in normal tissues (n = 18) (Figure 2(a)). To account for this difference, we analyzed the correlation of PCMT1 methylation and mRNA expression in breast cancer using the gene set cancer analysis (GSCA) database, which showed a significant negative correlation. Therefore, we speculate that the inconsistency of PCMT1 mRNA and protein expression in breast cancer stems from the inhibition of posttranslational modification and other processes in breast cancer, resulting in lower protein expression levels than normal tissues. The results are shown in Figure 2(d). There was no significant difference in the methylation level of PCMT1 in breast cancer (n = 793) and normal tissues (n = 97). In addition, we assessed the association between the expression of PCMT1 in breast cancer and tumor staging. The results showed that the expression level of PCMT1 was significantly correlated with the rising trend of breast cancer tumor stages (Figure 2(a)). Using the Human Protein Atlas database, we assessed the expression level of PCMT1 through immunohistochemical images of PCMT1 in clinical breast cancer.
Figure 7: Continued.
samples. The results showed that PCMT1 was overexpressed in tumor tissues (Figure 2(b)). Based on the CCLE database, we explored the expression level of PCMT1 in different breast cancer cell lines, and the expression level of PCMT1 was significantly different in different breast cancer cell lines. The heat map shows that PCMT1 is highly expressed in human breast ductal carcinoma cell lines, such as HCC1500, HCC1419, and EFM19 cell lines, while the expression is lowest in the HCC202 cell line (human breast primary ductal carcinoma cells) (Figure 2(c)).

We used bc-GenExMiner v4.7 to explore the relationship between PCMT1 and clinical pathological parameters of breast cancer. We observed that PCMT1 has a higher expression level in the age group over 51 years old ($P < 0.0001$). PCMT1 expressed higher in the negative progesterone receptor state ($P = 0.0042$). Compared with the human epidermal growth factor receptor 2 (HER2) negative group, the expression level of PCMT1 in the HER2 positive group was higher ($P = 0.0294$). In addition, for P53 status, we also observed that the mutant group had higher PCMT1 expression ($P = 0.0001$) (Table 1).

3.3. Gene Mutation and Immune Infiltration Level of PCMT1 in Breast Cancer. First, based on the existing ICGC/TCGA-Nature 2020 data set (2922 patients) [39] on the cbioPortal platform, we conducted a genome-wide pancancer analysis of PCMT1 gene mutations. The results showed that PCMT1 has the highest mutation frequency in soft tissue sarcoma, esophagogastric cancer, and breast cancer (Figure 3(a)). Then, we studied the PCMT1 gene mutation in breast cancer patients of TCGA Firehose Legacy ($n = 1108$ patients) [40]. Among the 1108 patients inquired, 19 (2%) patients had mutations in the PCMT1 gene, and the main types of mutations included missense mutations, increased gene duplication, and deletion mutations. The 86th amino acid in PCMT1 is prone to mutation, and the mutation type is a missense mutation (Figure 3(b)).

In addition, using the online server TIMER, we evaluated the relationship between the expression of PCMT1 and tumor immune infiltrating cell biomarkers, including B cells, CD8+ T cells, CD4+ T cells, macrophages, neutrophils, and dendritic cells. The results showed that PCMT1 was significantly positively correlated with tumor purity ($r = 0.132, P = 2.96e-05$), B cells ($r = 0.082, P = 1.06e-02$), CD8+ T cells ($r = 0.144, P = 5.96e-06$), macrophages ($r = 0.112, P = 4.16e-04$), neutrophils ($r = 0.105, P = 1.16e-03$), and dendritic cells ($r = 0.071, P = 2.76e-02$) and significantly negatively correlated with CD4+ T cells ($r = -0.069, P = 3.17e-02$) (Figure 3(c)). EPIC analysis showed that the expression of PCMT1 was significantly correlated with B cell, T cell CD4+, T cell CD8+, endothelial cell, and macrophage and NK cell (Figure 4(d)), and ssGSEA analysis also showed that PCMT1 expression was associated with various immune cell infiltration (Figures 4(a) and 4(c)). TIP analysis
showed that PCMT1 was significantly associated with breast cancer priming and activation (step 3), trafficking of immune cells to tumors (step 4), infiltration of immune cells into tumors (step 5), and recognition of cancer cells by T cells (step 6) (Figure 4(b)). These results support the involvement of PCMT1 in immune cell infiltration in breast cancer.

PCMT1 has a significant positive correlation with TMB \((r = 0.18, P = 4.29e-09)\) (Figure 3(d)), and a significant negative correlation with MSI \((r = -0.07, P = 0.025)\) (Figure 3(e)). We also evaluated the relationship between PCMT1 expression level and ICB response based on the TIDE algorithm. The results showed that high PCMT1 expression has a lower TIDE score, ICB curative effect is better, and survival after ICB treatment is longer (Figure 3(f)).

### 3.4. Analysis of PCMT1 Coexpressed Genes in Breast Cancer

To further explore the potential mechanism of PCMT1 in breast cancer, we used the LinkedOmics database to study the coexpression of PCMT1. The results showed that there was a significant positive correlation between PCMT1 and NUP43 \((r = 0.6322, P = 4.75e - 123)\) (Figures 5(a) and 5(b)). Studies have shown that the upregulation of NUP43 is associated with poorer OS in luminal A and HER2+ breast tumors [41]. Functional enrichment analysis was performed on the first 50 coexpressed genes, including biological process (BP), cellular component (CC), molecular function (MF), and Kyoto Encyclopedia of Genes and Genomes (KEGG). BP analysis showed that PCMT1-related coexpressed genes were mainly enriched in DNA-templated transcription, initiation, and translation. CC analysis showed that PCMT1-related coexpressed genes were mainly enriched in RNA polymerase II, holoenzymes, and mitochondrial matrix; MF analysis showed that PCMT1-related coexpressed genes were mainly enriched in isomerase activity; and KEGG analysis showed that PCMT1-related coexpressed genes were mainly enriched in basal transcription factors (Figures 5(c) and 5(d)).

### 3.5. Survival Analysis of PCMT1

Based on the online server Kaplan–Meier plotter, we evaluated the relationship between the expression level of PCMT1 and the survival of breast cancer patients to reveal the prognostic value of PCMT1 in...
breast cancer. The results showed that low expression of PCMT1 was significantly correlated with longer OS (HR = 1.38, 95% CI: 1.14–1.67, \( P = 0.00072 \)), RFS (HR = 1.23, 95% CI: 1.11–1.37, \( P = 4.9e-05 \)), and PPS (HR = 1.44, 95% CI: 1.14–1.82, \( P = 0.0023 \)) in breast cancer patients, but not significantly correlated with DMFS (HR = 0.87, 95% CI: 0.73–1.04, \( P = 0.12 \)) (Figure 6). This suggests that PCMT1 may have the function of a biomarker for the early diagnosis of breast cancer.

In addition, we analyzed the prognostic value of PCMT1 mRNA expression in different molecular subtypes according to the 2013 St. Gallen breast cancer criteria, including basal-like, luminal A, luminal B, and HER2+ subtypes. We evaluated the three probes (205202_at, 208857_s_at, 210156_s_at) in the Kaplan–Meier Plotter database. Survival outcomes include RFS, OS, DMFS, and PPS. We observed that high expression of PCMT1 was significantly associated with poor prognosis in the luminal A subtype (Table 2). The high expression of PCMT1 is significantly correlated with the worse RFS, OS, DMFS, and PPS of luminal A subtype.

3.6. The Relationship between the PCMT1 Gene and the Prognosis of Breast Cancer. Based on the raw count of the RNA sequencing data of 1097 breast cancer tumors downloaded from the TCGA data set and the corresponding clinical information, we studied the prognostic effect of the PCMT1 gene on breast cancer. We arranged the samples in order of PCMT1 expression level from high to bottom and used different grouping methods to analyze the prognostic differences of different groups. The results showed that PCMT1 gene expression and survival time in the TCGA data set were positively correlated with survival status (Figure 7(a)). The KM survival curve was drawn based on the TCGA data set and showed that high PCMT1 expression may be a risk factor for poor prognosis in breast cancer patients (HR: 1.92 (1.38, 2.67), \( P = 0.0001 \)) (Figure 7(b)). AUC can observe the effectiveness of PCMT1 as a prognostic biomarker. We drew the 1-year, 3-year, and 5-year ROC curves of the PCMT1 gene and calculated the AUC value. Among them, at 3 years, the AUC value was 0.683, indicating that PCMT1 has a certain diagnostic value.

Given that NUP43 is significantly associated with the prognosis of different subtypes of breast cancer, we analyzed the prognosis of PCMT1 in different subtypes of breast cancer, and the results showed that high expression of PCMT1 was significantly associated with poor prognosis in the luminal A subtype (see Supplementary Figure 3).

3.7. Construction and Verification of Nomogram Based on PCMT1. First, univariate Cox regression analysis was used to show that the expression of PCMT1 (HR: 2.245; CI: 1.54–3.28; \( P = 3e-05 \)), PT staging (HR: 1.70; CI: 1.30–2.20; \( P = 7e-05 \)), and PN staging (HR: 1.82; CI: 1.47–2.26; \( P < 0.0001 \)) was significantly related to prognosis. Multivariate Cox regression analysis showed that the expression of PCMT1 (HR: 2.46; CI: 1.62–3.73; \( P = 0.00002 \)), PT staging (HR: 1.48; CI: 1.10–1.99; \( P = 0.009 \)), and PN staging (HR: 1.70; CI: 1.34–2.16; \( P = 0.00001 \)) was also significant, indicating that the PCMT1 gene is a variable independent of other clinical factors.

We constructed nomograms using PCMT1 and independent clinical risk factors to provide a quantitative method for predicting disease-specific survival outcomes in breast cancer patients and luminal A subtypes. The 45° line represents the best prediction. In addition, we also analyzed the prediction efficiency of the nomogram, and the results showed that the C-index of the model was 0.746 (CI: 0.682–1) and 0.757 (CI: 0.689–1), respectively. It is shown that the model has good prediction accuracy for breast cancer prognosis (especially luminal A) (see Figure 8 and Supplementary Figure 4).

4. Discussion

Initial research showed the antiapoptotic effect of PCMT1 and speculated that its main mechanism is the ability to maintain the structural stability of some key antiapoptotic proteins through methylation and repair of some malfunctioning proteins [42]. The role of PCMT1 was subsequently proven in liver cancer. Studies have shown that PCMT1 is effectively regulated by the microRNA 15a/16-1 cluster and participates in cell apoptosis by protecting the structural stability and biological functions of BclxL (antiapoptotic mediator) from deamidation [43]. This proves that PCMT1 is involved in the regulation of hepatoma cell apoptosis. Zhao et al. constructed a random tumor transcriptome expression library to successfully create an A5 protein antigen targeting PCMT1 and showed a significant immunotherapy effect on S180 sarcoma [44]. Saito et al. proved that PCMT1 overexpression is an independent predictor of poor prognosis of lung adenocarcinoma through multivariate Cox risk regression analysis [45]. In addition, a study of PCMT1 in bladder cancer showed that PCMT1 regulates the migration and invasion of bladder cancer cells, promotes the occurrence and development of bladder cancer, and emphasizes that PCMT1 is an unfavorable prognostic biomarker for bladder cancer [15].

Although it has been confirmed that PCMT1 is involved in the occurrence and development of several cancers and its prognostic role has been emphasized, the role of PCMT1 in breast cancer has not been confirmed. In our study, we first performed a pancancer analysis of PCMT1, and the results showed that PCMT1 is highly expressed in a variety of cancers, including confirmed lung adenocarcinoma. The results of pancancer immune correlation evaluation show that, in addition to ACC, DLBC, and UVM, PCMT1 has varying degrees of correlation with the immune infiltrating cells of a variety of human cancers. Pancancer analysis also revealed that the expression of PCMT1 is significantly correlated with the prognosis of a variety of cancers. TMB is a biomarker that can help predict the patient’s response to immunotherapy [46]. We evaluated the correlation between TMB and PCMT1 expression in a variety of cancers, and the results showed that the expression of PCMT1 was significantly correlated with a variety of cancers, including ACC and BRCA. This finding suggests that PCMT1 may participate in the regulation of a variety of tumor-related signaling pathways and is significantly related to immune infiltration. Then,
we specifically studied the expression of PCMT1 in breast cancer. Compared with normal tissues, the PCMT1 gene is significantly more highly expressed in breast cancer, but its protein level is lower than that in normal tissues, revealing the influence of posttranslational modifications in breast cancer on PCMT1. In addition, the expression of PCMT1 is significantly related to the tumor stage of breast cancer, suggesting that it may be a prognostic marker of breast cancer. The analysis of breast cancer immune infiltration indicated that PCMT1 was significantly related to biomarkers of breast cancer immune infiltrating cells. The survival analysis of PCMT1 in breast cancer showed that the high expression of PCMT1 can lead to shorter OS, RFS, and PPS in breast cancer patients, suggesting that high expression of PCMT1 is significantly related to the poor prognosis of breast cancer. In addition, we constructed an ROC curve and a nomogram to observe the efficacy of PCMT1 as a prognostic biomarker. The results showed that the PCMT1 gene is a variable independent of other clinical factors and can guide the prognosis of breast cancer.

The limitation of our research is mainly reflected in the fact that all researches are based on the results of bioinformatics analysis. The potential biological mechanism of PCMT1 in breast cancer, its potential relationship with tumor immune escape, and its clinical role still need to be further studied. Our research results will provide certain reference value for further research on the role of PCMT1 in breast cancer.

5. Conclusion

In summary, our results show that the high expression of PCMT1 is significantly related to the poor prognosis of breast cancer, may be a potential biomarker of breast cancer, and is significantly related to the immune infiltration of breast cancer.

Data Availability

All data generated or analyzed during the current study are included in this published article and its supplementary information files.

Conflicts of Interest

The authors declare no conflict of interest.

Authors’ Contributions

JF, XL, and CL conceived the initial research project. JF and XL performed the statistical analysis and explained the data together with CL. CL wrote the first draft of the manuscript. All authors have read and agreed to the published version of the manuscript. Jufang Guo and Xuelian Du contributed equally to this work and should be considered co-first authors.

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Supplementary Materials

Supplementary Figure 1 The expression of immune checkpoint-related genes in different tumor tissues. Supplementary Figure 2 The correlation between tumor mutation burden and PCMT1 expression. Supplementary Figure 3 Relationship between PCMT1 and prognosis of different BRCA subtypes. Supplementary Figure 4 Construction of a predictive model of PCMT1 for luminal A isoforms. (Supplementary Materials)

References

[1] Y. Liang, H. Zhang, X. Song, and Q. Yang, “Metastatic heterogeneity of breast cancer: molecular mechanism and potential therapeutic targets,” Seminars in Cancer Biology, vol. 60, pp. 14–27, 2020.
[2] W. Hong and E. Dong, “The past, present and future of breast cancer research in China,” Cancer Letters, vol. 351, no. 1, pp. 1–5, 2014.
[3] R. L. Siegel, K. D. Miller, H. E. Fuchs, and A. Jemal, “Cancer statistics, 2021,” CA: a Cancer Journal for Clinicians, vol. 71, no. 1, pp. 7–33, 2021.
[4] S. Jiralerspong and P. J. Goodwin, “Obesity and Breast Cancer Prognosis: Evidence, Challenges, and Opportunities,” Journal of Clinical Oncology, vol. 34, no. 35, pp. 4203–4216, 2016.
[5] Y. S. Sun, Z. Zhao, Z. N. Yang et al., “Risk factors and predictions of breast cancer,” International Journal of Biological Sciences, vol. 13, no. 11, pp. 1387–1397, 2017.
[6] N. Brook, E. Brook, A. Dharmarajan, C. R. Dass, and A. Chan, “Breast cancer bone metastases: pathogenesis and therapeutic targets,” The International Journal of Biochemistry & Cell Biology, vol. 96, pp. 63–78, 2018.
[7] M. Rohanizadegan, “Analysis of circulating tumor DNA in breast cancer as a diagnostic and prognostic biomarker,” Cancer Genetics, vol. 228-229, pp. 159–168, 2018.
[8] H. Yang, S. Zhang, P. Liu et al., “Use of high-resolution full-field optical coherence tomography and dynamic cell imaging for rapid intraoperative diagnosis during breast cancer surgery,” Cancer, vol. 126, Supplement 16, pp. 3847–3856, 2020.
[9] R. Capecchi, I. Puxeddu, F. Pratesi, and P. Migliorini, “New biomarkers in SLE: from bench to bedside,” Rheumatology (Oxford, England), vol. 59, Supplement 5, pp. v12–v18, 2020.
[10] V. S. P. K. S. A. Jayanthi, A. B. Das, and U. Saxena, “Grade-specific diagnostic and prognostic biomarkers in breast cancer,” Genomics, vol. 112, no. 1, pp. 388–396, 2020.
[11] H. J. Wu and P. Y. Chu, “Recent discoveries of macromolecule- and cell-based biomarkers and therapeutic implications in breast cancer,” International Journal of Molecular Sciences, vol. 22, no. 2, p. 636, 2021.
[12] E. I. Ozay, S. Shanthalingam, J. A. Torres, B. A. Osborne, G. N. Tew, and L. M. Minter, “Protein kinase C theta modulates PCMT1 through hKRNPL to regulate FOX3 stability in regulatory T cells,” *Molecular Therapy*, vol. 28, no. 10, pp. 2220–2236, 2020.

[13] H. Zhao, F. Wang, J. Wang et al., “Maternal PCMT1 gene polymorphisms and the risk of neural tube defects in a Chinese population of Lviang high-risk area,” *Gene*, vol. 505, no. 2, pp. 340–344, 2012.

[14] F. Wang, J. Wang, J. Guo et al., “PCMT1 gene polymorphisms, maternal folate metabolism, and neural tube defects: a case-control study in a population with relatively low folate intake,” *Genes & Nutrition*, vol. 8, no. 6, pp. 581–587, 2013.

[15] L. Dong, Y. Li, D. Xue, and Y. Liu, “PCMT1 is an unfavorable predictor and functions as an oncogene in bladder cancer,” *IUBMB Life*, vol. 70, no. 4, pp. 291–299, 2018.

[16] P. Danaher, S. Warren, R. Lu et al., “Pan-cancer adaptive immunity resistance as defined by the tumor inflammation signature (TIS): results from the cancer genome atlas (TGCA),” *Journal for Immunotherapy of Cancer*, vol. 6, no. 1, pp. 63, 2018.

[17] F. G. Frost, P. F. Cherukuri, S. Milanovich, and C. F. Boerkoel, “Pan-cancer RNA-seq data stratifies tumours by some hallmarks of cancer,” *Journal of Cellular and Molecular Medicine*, vol. 24, no. 1, pp. 418–430, 2020.

[18] W. Lin, S. Wu, X. Chen et al., “Characterization of hypoxia signature to evaluate the tumor immune microenvironment and predict prognosis in glioma groups,” *Frontiers in Oncology*, vol. 10, p. 796, 2020.

[19] G. Sturm, F. Finotello, F. Pettitprez et al., “Comprehensive evaluation of transcriptome-based cell-type quantification methods for immuno-oncology,” *Bioinformatics*, vol. 35, no. 14, pp. 436–445, 2019.

[20] T. Li, J. Fu, Z. Zeng et al., “TIMER2.0 for analysis of tumor-infiltrating immune cells,” *Nucleic Acids Research*, vol. 48, no. W1, pp. W509–W514, 2020.

[21] D. Zeng, M. Li, R. Zhou et al., “Tumor microenvironment characterization in gastric cancer identifies prognostic and immunotherapeutically relevant gene signatures,” *Cancer Immunology Research*, vol. 7, no. 5, pp. 737–750, 2019.

[22] V. Thorsson, D. L. Gibbs, S. D. Brown et al., “The immune landscape of cancer,” *Immunity*, vol. 48, no. 4, pp. 812–830.e14, 2018.

[23] D. S. Chandrashekar, B. Bashel, S. A. H. Balasubramanyam et al., “UALCAN: a portal for facilitating tumor subgroup gene expression and survival analyses,” *Neoplasia*, vol. 19, no. 8, pp. 649–658, 2017.

[24] M. Uhlen, C. Zhang, S. Lee et al., “A pathology atlas of the human cancer transcriptome,” *Science*, vol. 357, no. 6352, article eaan2507, 2017.

[25] D. P. Nusinow, J. Szpyt, M. Ghandi et al., “Quantitative proteomics of the cancer cell line encyclopedia,” *Cell*, vol. 180, no. 2, pp. 387–402.e16, 2020.

[26] P. Jézéquel, W. Gourraud, F. Ben Azzouz et al., “bc-Gen ExMiner 4.5: new mining module computes breast cancer differential gene expression analyses,” *Database*, vol. 2021, article baab007, 2021.

[27] J. Gao, B. A. Aksoy, U. Dogrusoz et al., “Integrative analysis of complex cancer genomics and clinical profiles using the cBioPortal,” *Science Signaling*, vol. 6, no. 269, p. pii, 2013.

[28] T. Li, J. Fan, B. Wang et al., “TIMER: a web server for comprehensive analysis of tumor-infiltrating immune cells,” *Cancer Research*, vol. 77, no. 21, pp. e108–e110, 2017.

[29] L. Xu, C. Deng, B. Pang et al., “TIP: a web server for resolving tumor immunophenotype profiling,” *Cancer Research*, vol. 78, no. 23, pp. 6575–6580, 2018.

[30] P. Jiang, S. Gu, D. Pan et al., “Signatures of T cell dysfunction and exclusion predict cancer immunotherapy response,” *Nature Medicine*, vol. 24, no. 10, pp. 1550–1558, 2018.

[31] S. V. Vasaikar, P. Straub, J. Wang, and B. Zhang, “LinkedOmics: analyzing multi-omics data within and across 32 cancer types,” *Nucleic Acids Research*, vol. 46, no. D1, pp. D956–D963, 2018.

[32] G. Bindea, B. Mlecnik, H. Hackl et al., “ClueGO: a Cytoscape plug-in to decipher functionally grouped gene ontology and pathway annotation networks,” *Bioinformatics*, vol. 25, no. 8, pp. 1091–1093, 2009.

[33] G. Bindea, J. Galon, and B. Mlecnik, “CluePedia Cytoscape plugin: pathway insights using integrated experimental and in silico data,” *Bioinformatics*, vol. 29, no. 5, pp. 661–663, 2013.

[34] P. Shannon, A. Markiel, O. Ozier et al., “Cytoscape: a software environment for integrated models of biomolecular interaction networks,” *Genome Research*, vol. 13, no. 11, pp. 2498–2504, 2003.

[35] Y. Zhou, B. Zhou, L. Pache et al., “Metascape provides a biologist-oriented resource for the analysis of systems-level datasets,” *Nature Communications*, vol. 10, no. 1, pp. 1523, 2019.

[36] B. Győrffy, “Survival analysis across the entire transcriptome identifies biomarkers with the highest prognostic power in breast cancer,” *Computational and Structural Biotechnology Journal*, vol. 19, pp. 4101–4109, 2021.

[37] C. Zhang, Y. Zheng, X. Li, X. Hu, F. Qi, and J. Luo, “Genome-wide mutation profiling and related risk signature for prognosis of papillary renal cell carcinoma,” *Ann Transl Med.*, vol. 7, no. 18, p. 427, 2019.

[38] A. Iasonos, D. Schrag, G. V. Raj, and K. S. Panagagua, “How to build and interpret a nomogram for cancer prognosis,” *Journal of Clinical Oncology*, vol. 26, no. 8, pp. 1364–1370, 2008.

[39] ICGC/TCGA Pan-Cancer Analysis of Whole Genomes Consortium, “Pan-cancer analysis of whole genomes,” *Nature*, vol. 578, no. 7793, pp. 82–83, 2019.

[40] Cancer Genome Atlas Network, “Comprehensive molecular portraits of human breast tumours,” *Nature*, vol. 490, no. 7418, pp. 61–70, 2012.

[41] C. Tian, S. Zhou, and C. Yi, “High NUP43 expression might independently predict poor overall survival in luminal A and in HER2+ breast cancer,” *Future Oncology*, vol. 14, no. 15, pp. 1431–1442, 2018.

[42] I. Sambri, R. Capasso, P. Pucci, A. F. Perna, and D. Ingrosso, “The microRNA 15a/16-1 cluster down-regulates protein repair isoaspartyl methyltransferase in hepatoma cells: implications for apoptosis regulation,” *The Journal of Biological Chemistry*, vol. 286, no. 51, pp. 43690–43700, 2011.
[43] M. Amer, M. Elhefnawi, E. El-Ahwany et al., “Hsa-miR-195 targets PCMT1 in hepatocellular carcinoma that increases tumor life span,” Tumour Biology, vol. 35, no. 11, pp. 11301–11309, 2014.

[44] H. Zhao, X. Zhao, P. Du, and G. Qi, “Construction of random tumor transcriptome expression library for creating and selecting novel tumor antigens,” Tumour Biology, vol. 37, no. 9, pp. 12877–12887, 2016.

[45] H. Saito, M. Yamashita, M. Ogasawara et al., “Chaperone protein L-isoaspartate (D-aspartyl) O-methyltransferase as a novel predictor of poor prognosis in lung adenocarcinoma,” Human Pathology, vol. 50, pp. 1–10, 2016.

[46] T. A. Chan, M. Yarchoan, E. Jaffee et al., “Development of tumor mutation burden as an immunotherapy biomarker: utility for the oncology clinic,” Annals of Oncology, vol. 30, no. 1, pp. 44–56, 2019.