

## Analysis

## The expression and prognostic value of CCL19 in breast cancer

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© The Author(s) 2025 **OPEN****Abstract**

**Background** Breast cancer ranks as the foremost cause of cancer-related mortality among women globally. Timely diagnosis stands as the most effective approach in mitigating breast cancer mortality rates. There exists a close relationship between immune processes and tumorigenesis. This study aims to elucidate the immune mechanisms and potential biomarkers associated with breast cancer using bioinformatics techniques.

**Objective** Initially, differentially expressed genes were identified through consensus analysis of invasive breast cancer (BRCA) samples sourced from The Cancer Genome Atlas (TCGA) database, focusing on immunotherapy response. Subsequently, protein–protein interaction (PPI) networks, Gene Ontology (GO), and Kyoto Encyclopedia of Genes and Genomes (KEGG) analyses were employed to refine the selection of potential genes. Lastly, expression and prognostic analyses of hub genes were conducted to identify reliable key genes, with a focus on *CCL19*. Immunohistochemistry (IHC) was employed to assess the differential expression of *CCL19* in both tumor and adjacent breast tissue samples. Additionally, protein correlation analysis, signaling pathway analysis, immune infiltration analysis, gene co-expression analysis, and drug sensitivity analysis of *CCL19* were conducted to investigate its pathological clinical features and potential biological functions.

**Results** *CCL19* expression exhibited a significant increase in breast cancer. Elevated *CCL19* expression correlates with advanced tumor stage and indicates a favorable prognosis in breast cancer. *CCL19* expression correlates with the abundance of diverse tumor-infiltrating immune cells (TIICs). *CCL19* exhibits a positive correlation with the expression of the majority of immune-related genes. Enrichment analysis revealed the involvement of *CCL19* in immune-related pathways and the PI3K-Akt signaling pathway. These findings suggest that *CCL19* may influence breast cancer prognosis through immune infiltration. Patients exhibiting high *CCL19* expression demonstrated more favorable responses to immunotherapy.

**Conclusion** Breast cancer demonstrates overexpression of *CCL19*. *CCL19* holds promise as a biomarker for forecasting breast cancer prognosis and as a potential therapeutic target.

**Keywords** *CCL19* · Breast cancer · Biomarker · Immune infiltration

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## 1 Introduction

Breast cancer ranks as the most prevalent malignancy among women globally. The incidence of breast cancer has been rising in recent years. In 2020, roughly 2.3 million new cases of breast cancer were diagnosed globally among women. Breast cancer is the leading cause of cancer-related mortality among women, representing a significant threat to women's health [1]. It is widely recognized that tumor cells are regulated by the tumor microenvironment (TME). The interactions between tumor cells and the TME complicate the process of tumorigenesis, making it challenging to elucidate [2]. Thus, a comprehensive investigation into the expression levels of cancer-related genes is crucial for the diagnosis, treatment, and prognosis of breast cancer.

*CCL19*, located on human chromosome 17, belongs to the chemokine family. *CCL19* is produced by lymph nodes, the spleen, and dendritic cells and exhibits abundant expression in the thymus and lymph nodes [3]. *CCL19* plays a significant role in immune response, particularly in the transportation of T cells, NK cells, and dendritic cells (DCs). *CCL19* mediates tumor progression via its ligand CCR7 [4]. A growing body of literature has documented differential expression of *CCL19* in various malignant tumors, including gastric cancer, lung cancer, cervical cancer, and colorectal cancer [5–8]. This suggests that *CCL19* may play a potential role in tumorigenesis. *CCL19* has been shown to suppress tumors in breast, gastric, and colorectal cancer (CRC) [5, 7, 9]. Conversely, overexpression of *CCL19* has been associated with tumor-promoting effects in small cell lung cancer (SCLC) and cervical cancer [6, 8]. Several studies have identified *CCL19* as a potential therapeutic target for cancer. Injection of *CCL19* into mouse models of both lung cancer and colorectal cancer has led to tumor shrinkage [5, 10]. However, research on the specific mechanisms and therapeutic applications of *CCL19* in breast cancer remains limited.

Consequently, we employed a range of methodologies, including gene expression analysis, protein correlation analysis, signaling pathway analysis, prognostic analysis, gene co-expression analysis, and drug sensitivity analysis, to comprehensively investigate the role of *CCL19* in breast cancer. The expression of *CCL19* was analyzed in conjunction with tumor-infiltrating lymphocytes (TILs) and immunomodulatory factors. Our results indicate a close association between *CCL19* and breast cancer immunity. *CCL19* may serve as a significant predictive biomarker for breast cancer prognosis and could represent a novel therapeutic target.

## 2 Methods

### 2.1 Ethical approval

This study was approved by the institutional research ethics committee of the Second Affiliated Hospital of Anhui Medical University (Approval No. AHMU2 ND-202178). All 101 patients were pathologically confirmed breast cancer patients who treated at the Second Affiliated Hospital of Anhui Medical University from January 2017 to December 2021. Patients with comorbidity of other malignancies were excluded. All selected patients had no prior chemotherapy, radiotherapy, or targeted therapy.

### 2.2 Data collection

GSE108474 data were downloaded from the Gene Expression Omnibus (GEO, <http://www.ncbi.nlm>). The GSE103668 (Triple Negative Breast cancer) dataset was used to predict immunotherapy response. The transcribed values are transformed log2 using the "limma" package in R. All data involved in this study were obtained and applied in accordance with the database Publication Guidelines and data access policies.

### 2.3 Identification of DEGs

21 breast cancer samples from GSE108474 data were divided into two groups based on whether there was a response to immunotherapy. "Limma" of R software was used to determine the differentially expressed genes (DEGs) between different clusters [11]. Adjusted p value < 0.05, |logF<sub>c</sub>| > 1 was defined as the screening threshold for differential mRNA expression. The heatmaps were all visualized by the R package heatmap. The SRTING database (cn.string-db.org) was

used to construct the PPI network. Data obtained from the SRTING database were used to identify 20 hub genes with the cytoHubba plug-in of Cytoscape software [12] and combined into a visual network. To further confirm the potential functions of DEGs, enrichment analysis was performed. The “Cluster Profiler” package in R was used to analyze the GO function and the KEGG pathway [13].

## 2.4 CCL19 expression analysis

Gene expression data and clinical information of breast cancer were downloaded from TCGA. Additional pan-cancer datasets based on the GEO cohort were obtained from the GENT2 database. The differences of *CCL19* mRNA expression levels of tumor tissues and normal tissues among various tumor types was analyzed. Next, the clinical data downloaded were used to divide the samples into two groups based on the expression level of *CCL19*. The differences in clinical data between the two groups were analyzed separately.

## 2.5 Prognosis analysis

The data from TCGA database was used to assess the difference in survival between the *CCL19* high group and *CCL19* low group. Overall survival (OS) and disease free survival (DFS) were used to assess the prognostic value of *CCL19* in breast cancer. Multivariate Cox regression analysis was performed using the “survival” and “rms” packages in R software. The analysis included factors such as *CCL19*, age, gender, and tumor stage to construct a Cox proportional hazards model. The “regplot” package was used to visualize the model as a nomogram, and a calibration curve for the prediction model was generated simultaneously.

## 2.6 Genetic mutation analysis

The genetic mutation characteristics of *CCL19* were studied through the CBioPortal database ([www.cbioportal.org](http://www.cbioportal.org)) [14]. Further, based on the *CCL19* expression data from TCGA, the correlation of tumor mutational burden (TMB)/microsatellite instability (MSI) and *CCL19* expression was described by using Spearman analysis. The above figure was implemented by the R package “ggstatsplot”.

## 2.7 Immune infiltration analysis

CIBERSORT [15] was used to calculate the degree of immune cell infiltration in the BRCA queue in TCGA. Wilcoxon signed rank test was applied to analyze the differences in tumor infiltrating immune cells between different groups. The ESTIMATE algorithm was used to calculate the immune and matrix scores between two groups [16]. We use TIMER 2.0 database ([timer.cistrome.org](http://timer.cistrome.org)) to study the association of *CCL19* expression with immune cell infiltration [17]. Depmap (<https://depmap.org/portal/>) was used to obtain genome information of breast cancer cell lines and sensitivity to genetic and small molecule perturbations.

## 2.8 Gene co-expression analysis

Oncomine database ([www.oncomine.org](http://www.oncomine.org)) is a tumor microarray database, which was used for gene co-expression analysis of *CCL19* in this study. *CCL19* was subjected to gene co-expression analysis including immune activator, immunosuppressant, chemokine, chemokine receptor and MHC genes through the TIBIS database ([cis.hku.hk/TISIDB/index.php](http://cis.hku.hk/TISIDB/index.php)).

## 2.9 PPI network and gene enrichment analysis

SRTING database ([cn.string-db.org](http://cn.string-db.org)) was used to conduct an online search for the mutual aid relationship of *CCL19* proteins [18]. The obtained results were visualized with Cytoscape [12]. The R package “enrich plot” was used to study the enrichment of *CCL19* in signaling pathways. GO and KEGG were used to assess the potential function of *CCL19* through R package ‘clusterProfiler’. The R package “ggplot2” was used for plotting.

## 2.10 Drug susceptibility analysis

Considering the role of immune checkpoint inhibitors in immunotherapy, the correlation between *CCL19* and immune checkpoints was analyzed. Furthermore, the response of immunotherapy between the low expression and high expression *CCL19* groups was predicted. The immunophenotype score (IPS) reflects the efficacy of immunotherapy. The higher the IPS score, the stronger the immunogenicity. The IPS of TCGA-BRCA patients was derived from the Cancer Immunome Atlas (TCIA) (<https://TCIA.at/home>).

Genomics of Cancer Drug Sensitivity (GDSC) ([www.cancerrxgene.org](http://www.cancerrxgene.org)), the largest pharmacogenomics database, was used to predict chemotherapy response of 1101 breast cancer samples from TCGA [19]. The prediction process is implemented by the R package “pRRophetic” [20]. The mRNA expression data and drug sensitivity data were combined and subjected to Pearson correlation analysis to obtain the correlation of *CCL19* expression with IC50 value of the drugs. P-values were adjusted by false discovery rate (FDR).

## 2.11 IHC

Specimens from patients were immersed in methanol solution and embedded in paraffin. Anti-*CCL19* (1:200; Abcam) antibody and Anti-CD8 (1:300; Abcam) antibody were used for immunohistochemical staining. Paraffin blocks were cut into 3  $\mu$ m slices and placed in a 57 °C environment for 90 min to dry. Slices were placed in xylene and a range of concentrations of alcohol. Slices were subsequently boiled in a citrate solution. Endogenous hydrogen peroxide activity was blocked by hydrogen peroxide (0.3%). Slices were incubated with primary antibody overnight at 4 °C. After removal, wash three with phosphate-buffered saline (PBS) and incubate with secondary antibody for 30 min at 37 °C. Sections were stained in 3,3-diaminobenzidine tetrahydrochloride (DAB) and hematoxylin. The staining results were scored by two pathologists, who were unaware of the sample's information. Positive reactions were defined as brown staining. The IHC score is calculated by the staining intensity (0, negative; 1, weak; 2, moderate; 3, strong) multiplied by the positive area (0, less than 5%; 1, 5% to 25%; 2, 26% to 50%; 3, 51% to 75%; 4, greater than 75%). The full score is 12 points, where 0–7 points are considered low expression and 8–12 points are considered high expression [21].

## 2.12 Cell culture

Human breast cancer cell lines (MCF-7 and MDA-MB-231) were bought from the American Type Culture Collection (ATCC). All cells were cultured in DMEM medium with 10% fetal bovine serum and 100  $\times$  penicillin–streptomycin solution. All cells were cultured at 5% CO<sub>2</sub> and 37 °C. Recombinant Human *CCL19* (JN0318, Baiaoleibo, Beijing) was added to the cell line and continued to culture.

## 2.13 Cell proliferation assay

Breast cancer cells were inoculated into 96-well plates. Cell viability was calculated by CCK-8 system after 24, 48 and 72 h. The optical density (OD) value of each hole was measured at 450 nm (BioTek USA).

## 2.14 Colony formation assay

Breast cancer cells were cultured in 6-well plates for 2 weeks. Culture 500 cells per well until colony formation. The formed colonies were fixed with 4% paraformaldehyde (PFA) for 30 min and stained with 0.5% crystal violet for 5 min. The number of clones formed (more than 50 cells per clone) was manually measured and photographed.

## 2.15 Transwell

For invasion assay, the top chamber was pre-coated with Matrigel. Next, the cells were added culturing in a serum-free medium. In the meantime, RPMI-1640 medium with 10% FBS was put in the lower chamber. After that, cells that remained in the upper chamber were removed. Invaded cells were finally stained with crystal violet and observed.

For migration assay, the upper chamber was not coated with Matrigel, and the remaining procedures were similar to those in the invasion assay.

## 2.16 Statistical analysis

All the above statistical analysis and R packages were performed using R software (version 4.0.2) ([www.Rproject.org](http://www.Rproject.org)) for statistical analysis.  $P < 0.05$  indicates statistical significance.

## 3 Results

### 3.1 Identification of DEGs

The flow chart of this study is shown in Fig. 1. BRCA samples from GSE103668 were divided into two groups by the response to immunotherapy. 583 DEGs in total were obtained by screening the two groups, of which 581 genes were up-regulated and 2 genes were down-regulated (Fig. 2A). To further investigate the biological functions and underlying mechanisms of DEGs in breast cancer, GO enrichment analysis and KEGG pathway analysis were used. GO analysis showed that DEGs were mainly enriched in negative regulation of immune system process, regulation of cell–cell adhesion, leukocyte cell–cell adhesion, and T cell activation (Fig. 2B). KEGG analysis showed that DEGs were mainly involved in chemokine signaling pathway, cytokine-cytokine receptor interaction, and cell adhesion molecules (Fig. 2C).

### 3.2 The Expression Level and Prognostic Value of the Hub Gene

DEGs PPI network graphs are constructed. Cytoscape's plugin cytoHubba was used to seek for the most important nodes, and finally *PTPRC*, *CST7*, *C1S*, *CD2*, *IL7R*, *CD3E*, *CD3D*, *GZMA*, *CCL5*, *LYZ*, *CXCL9*, *NKG7*, *GBP5*, *CXCL11*, *CD79A*, *CXCL10*, *JCHAIN*, *CCL19*, *CCL21*, *CXCL13* were identified as hub genes (Fig. 2D). To determine the diagnostic significance and prognostic value of hub genes, the GEPIA database was used to validate the expression (Supplementary Fig. 1) and prognosis (Supplementary Fig. 2) of 20 genes in breast cancer. The above results show that *CCL19* is differentially expressed in breast cancer tissues. Moreover, the expression of *CCL19* is related to the OS of patients. Previous studies have shown that *CCL19* exists in the tumor microenvironment of breast cancer and plays an important role [3]. Therefore, *CCL19* was selected as the hub gene in this study.

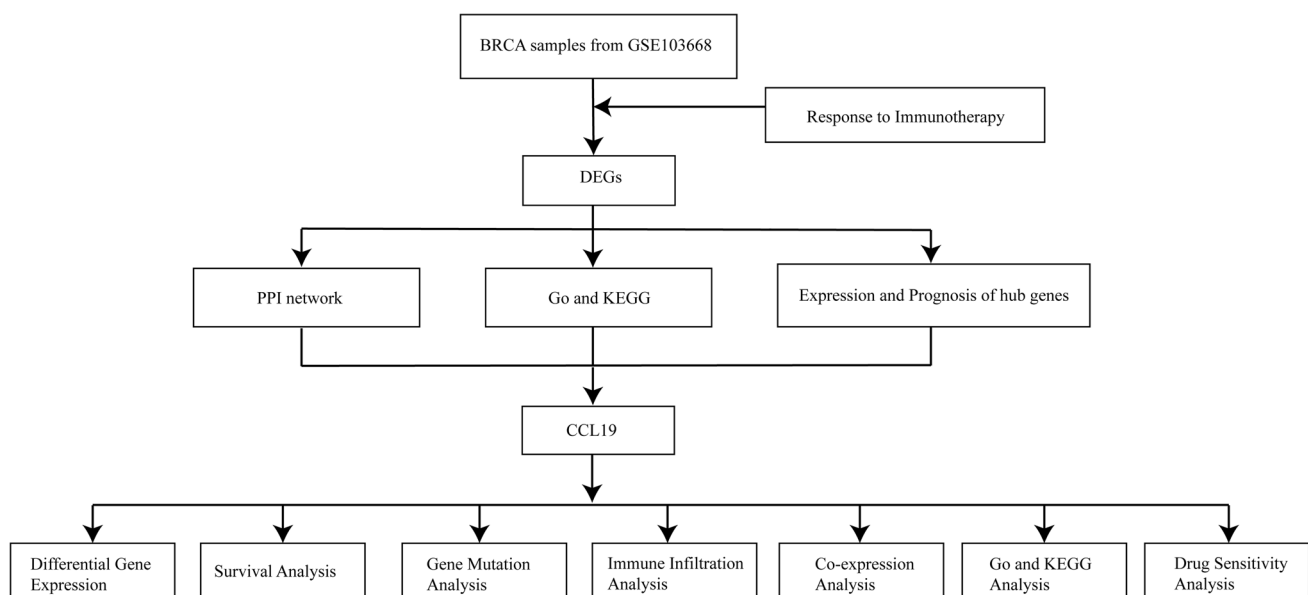
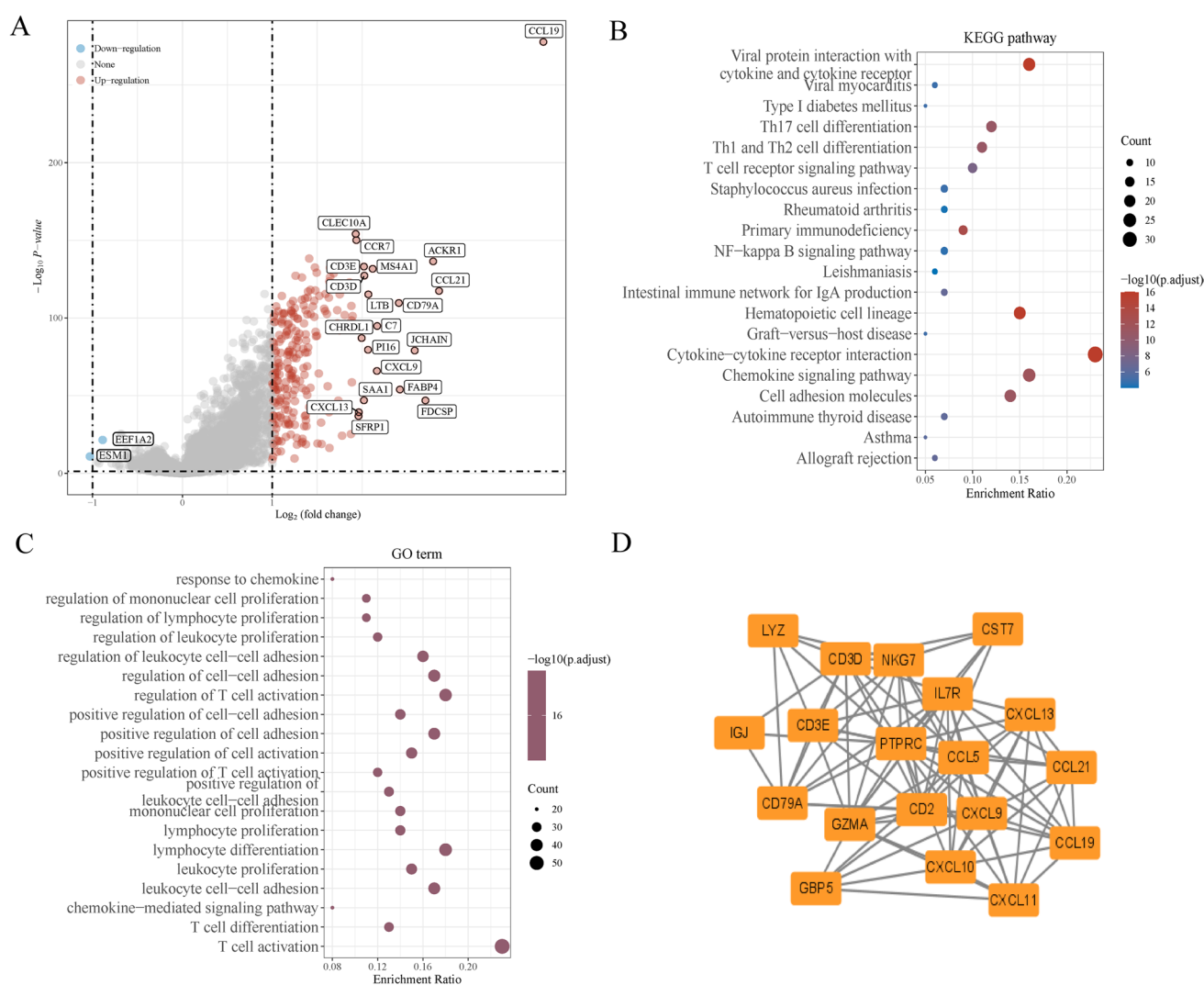


Fig. 1 Flowchart of this study

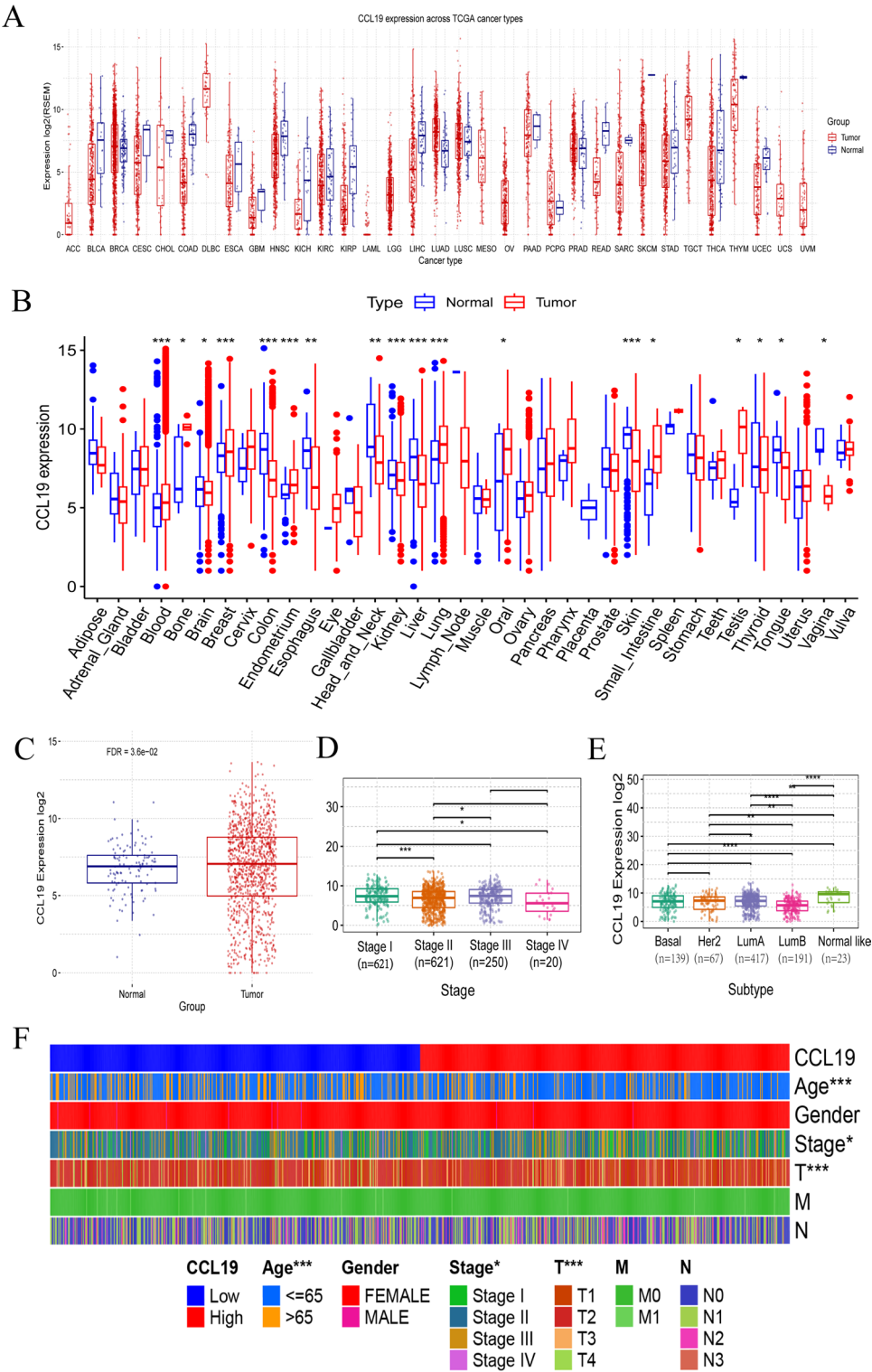


**Fig. 2** Analysis of DEGs based on immunotherapy response. **A** DEGs volcano map analysis; **B** KEGG pathway of DEGs; **C** GO analysis of DEGs; **D** Hub gene was selected from PPI network through Cytoscape plug-in Cytohubba

### 3.3 Expression of *CCL19* in tumor tissue

First, the mRNA expression level of *CCL19* in different tumor tissues was analyzed. Evaluation of *CCL19* mRNA expression in pan-cancer by TCGA database showed that *CCL19* expression was lower in bladder Urothelial Carcinoma (BLCA), cholangio carcinoma (CHOL), colon adenocarcinoma (COAD), esophageal carcinoma (ESCA), head and Neck squamous cell carcinoma (HNSC), kidney renal clear cell carcinoma (KIRC), prostate adenocarcinoma (PRAD), rectum adenocarcinoma (READ), thyroid carcinoma (THCA), uterine Corpus Endometrial Carcinoma (UCEC), and higher in BRCA (Fig. 3A). Data from the GEO database also support the above results (Fig. 3B). Furthermore, analysis of differential expression of *CCL19* based on cancer and cancer adjacent tissue samples in the TCGA database showed that *CCL19* expression was significantly higher in breast cancer tumor tissues compared with normal tissues (Fig. 3C). *CCL19* is differentially expressed between different molecular types of breast cancer, including triple-negative, HER2-overexpressing, Luminal A, and Luminal B (Fig. 3D). *CCL19* expression was assessed in different cancer stages, and *CCL19* expression was statistically different between tumor stages I and II, II and III, and II and V (Fig. 3E). The data were divided into two groups according to the expression level of *CCL19*, and the differences in clinical data between the two groups were analyzed. There were significant differences in age, tumor stage, and T classification between the two groups (Fig. 3F).

**Fig. 3** Differential expression of *CCL19*. **A** Expression of *CCL19* in different tumors from TCGA database; **B** Expression of *CCL19* in different tumors from GEO database; **C** Expression of *CCL19* in tumor tissues and normal tissues of breast cancer from TCGA database; **D** Expression of *CCL19* in different subtypes of breast cancer from TCGA database; **E** Expression of *CCL19* in different stages of breast cancer from TCGA database; **F** Correlation analysis between *CCL19* and different pathological features in breast cancer



3.4 Prognosis Value of *CCL19* in breast cancer

To further inspect the relationship between *CCL19* expression and breast cancer prognosis and survival, we divided breast cancer samples into high and low expression group according to the median value of *CCL19* expression level.

By analyzing survival data with the Kaplan–Meier curve, we noticed that *CCL19* expression level was significantly associated with better OS (Fig. 4A, B).

Univariate and multivariate Cox regression analyses were conducted to determine independent prognostic factors. Univariate Cox regression analysis showed that age, stage, T, N, and *CCL19* were closely related to survival rate (Fig. 4C). The above-mentioned influencing factors were incorporated into the multi-factor COX analysis. The results showed that *CCL19* ( $P = 0.017$ , HR = 0.911, 95%CI 0.844–0.983) was also significantly correlated with survival rate (Fig. 4D). In summary, the expression of the *CCL19* can be preliminarily determined that the volume of *CCL19* is related to the prognosis of patients. A prognostic nomogram including *CCL19* was constructed, which also included age, gender, TNM, Grade, and Stage. Among them, age has a higher influence, followed by *CCL19* expression (Fig. 4E). The calibration curve obtained simultaneously with the nomogram. The curves of 1-year, 3-year, and 5-year prediction values are basically distributed along the diagonal line (Fig. 4F), indicating that the model has good predictive performance.

### 3.5 Genomic mutation of *CCL19* in breast cancer

Since cancer is associated with Genomic mutations, we explored genomic alterations in *CCL19*. *CCL19* had different alterations frequencies in different breast cancer subtypes, including 1.39% (breast invasive ductal carcinoma) and 0.84% (breast mixed ductal and lobular carcinoma) (Supplementary Fig. 3 A). The "Mutations" modules of the cBioPortal database showed that the alterations frequency of *CCL19* was 1.1%, including amplification, deep deletion, and missense mutation (Supplementary Fig. 3B). In total, only one sample of *CCL19* had a missense mutation. Supplementary Fig. 3 C shows the location of the alterations in *CCL19*. Amplification is the most common type of *CCL19* alterations. We further explored the relationship between *CCL19* expression and TMB (tumor mutational burden) and MSI (microsatellite instability) of breast cancer in TCGA. We noticed that the expression of *CCL19* was negatively correlated with TMB and MSI (Supplementary Fig. 3D and E).

### 3.6 PPI network and enrichment analysis of *CCL19* in breast cancer

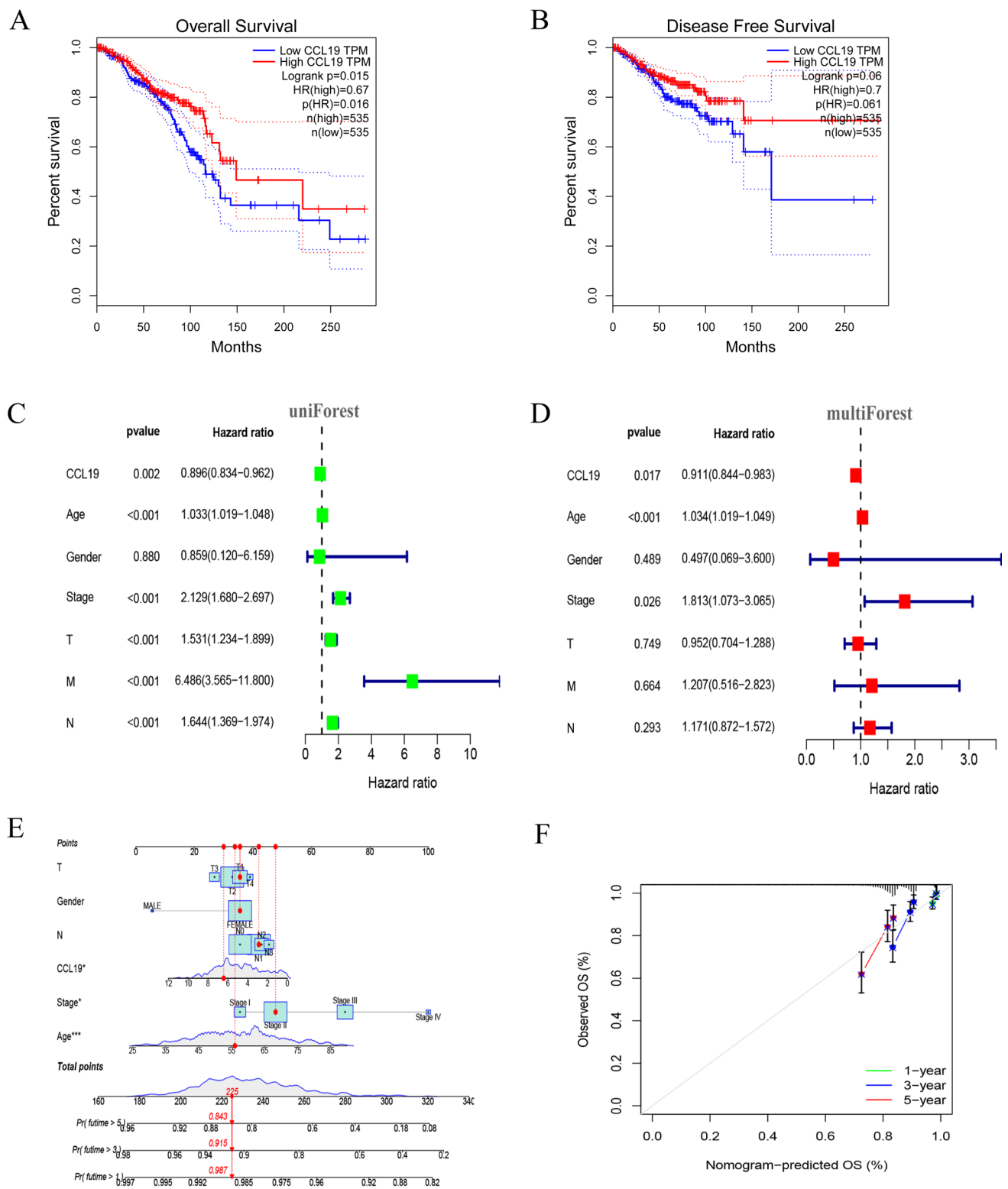
The genes significantly associated with *CCL19* were identified through the dataset from TCGA (Supplementary Table 1). Ten co-expressed genes significantly positively or negatively correlated with *CCL19* were shown in the circular graph (Fig. 5A). According to the median expression level of *CCL19*, breast cancer samples were divided into high and low expression group. Next, we compared the DEG analysis between the two groups.  $|\log_2(\text{multiple change})| > 1$ , and adjusted for  $p < 0.05$ . As the result, 1386 genes were identified as DEG (Supplementary Table S2). The top 50 upregulated genes and top 50 downregulated genes were shown in the heatmap (Fig. 5B).

The obtained genes were subjected to GO and KEGG analysis using the "clusterProfiler" package. The gene set with the highest credibility was selected to draw a ring diagram (Fig. 5C). The interaction between *CCL19* and its functional related proteins was shown in Supplementary Fig. 4 A-B. GO analysis found that *CCL19* is involved in Leukocyte mediated immunity, positive regulation of cell activation, and antigen binding (Supplementary Fig. 5 A, Supplementary Table 3). The results of KEGG showed that *CCL19* expression was associated with Cytokine-cytokine receptor interaction, viral protein interaction with cytokine and cytokine-receptor and Primary immunity (Supplementary Fig. 5B, Supplementary Table 4). GSEA was used to determine the signaling pathway of different activated *CCL19* related genes in breast cancer. The results showed that the immune related gene set was significantly enriched in *CCL19* related DEG (Fig. 5D).

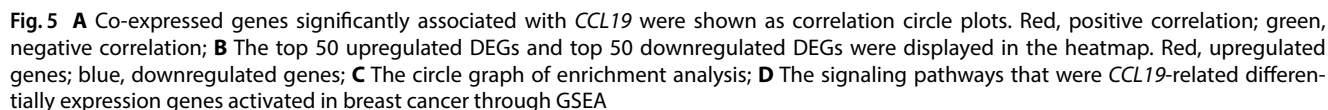
### 3.7 The Relationship between *CCL19* Expression and Immune Cell Infiltration in breast cancer

CIBERSORT calculation was used to determine the proportion of 22 immune cells infiltrated in the tissues of each breast cancer patient. The sample was divided into high and low expression groups according to the expression level of the target gene *CCL19*. The "limma" package was used to compare the differences in immune cells between the high and low expression groups. Figure 6A shows the distribution of immune infiltrating cells in the two groups. The correlation between *CCL19* expression and immune cell expression was visualized as a Lollipop plot (Fig. 6B). The high expression group of *CCL19* showed an increase in CD8 + T cells and M1 macrophages (Fig. 6C, D). The expression of *CCL19* was significantly negatively correlated with M0 macrophages and M2 macrophages (Fig. 6E, F).

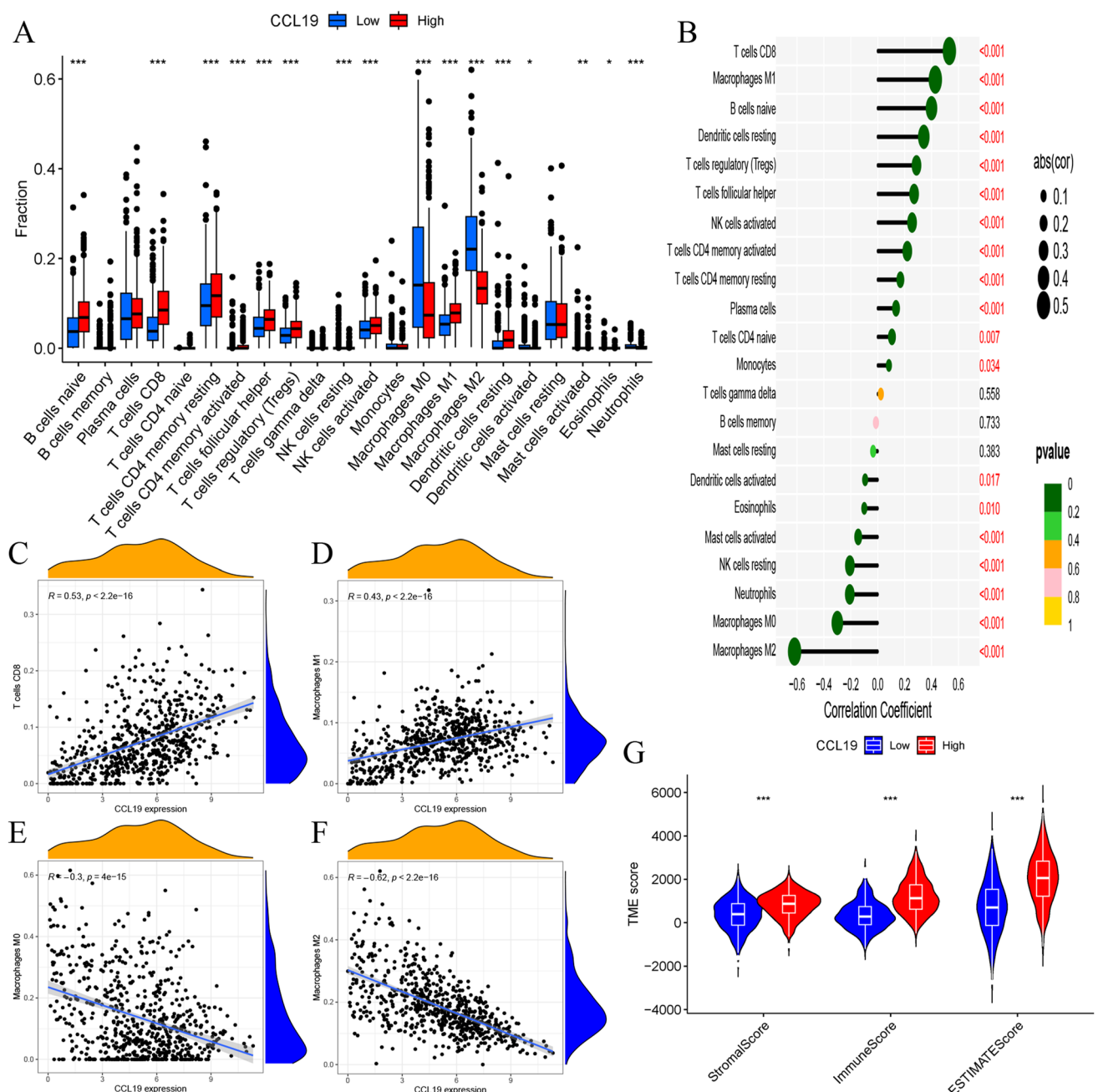
The correlation between *CCL19* expression and immune cell infiltration in pan cancer of the TIMER database is shown in Supplementary Fig. 6. XCELL, TIMER, CIBERSORT-ABS, QUANTISEQ, MCPOUNTER, and EPIC were integrated to compare the correlation between different groups of immune cells, indicating differences in immune



**Fig. 4** Survival analysis curve of *CCL19* in breast cancer. **A, B** OS and DFS in breast cancer patients through Kaplan–Meier curve; **C** Univariate Cox regression analysis of *CCL19* in breast cancer; **D** Multivariate Cox regression analysis of *CCL19* in breast cancer; **E** The prognostic nomogram with *CCL19* and clinical variables; **F** The 1-, 3- and 5-year calibration plots demonstrated the performance of the nomogram



status between low and high *CCL19* expression groups (Supplementary Fig. 7A). After adjusting the purity of the tumor, further analysis of the correlation between *CCL19* and the abundance of cancer immune cells revealed that *CCL19* expression was negatively correlated with B cells, CD4-T naïve cells, CD8-T cells, CD4-T memory, CD4-T effector memory and Myeloid DC cells (Supplementary Fig. 7B-G). Next, estimation algorithms were used to investigate the correlation immune score and matrix score between the two groups. We found that the immune score and matrix score of the high *CCL19* expression group were higher than those of the low *CCL19* expression group (Fig. 6G), which proved that the expression level of *CCL19* could affect the immune activity of breast TME in breast cancer.



**Fig. 6** **A** The proportion of 22 immune cells infiltrating in low and high *CCL19* expression groups; **B** Relationships between the expression of *CCL19* and 22 types of immune infiltration cells; **C–F** The relationships between the expression of *CCL19* and CD8 + T cells, M0 macrophages, M1 macrophages and M2 macrophages; **G** ESTIMATE algorithm was used to investigate the correlation between the two groups in Immune scores and Stromal scores

To understand the effect of *CCL19* knockout on the viability of different breast cancer cells, we explored the *CCL19* loss screening from Depmap. The Gene Effects value of *CCL19* in breast cancer cell lines is less than 0, which means that *CCL19* will inhibit cell proliferation (Supplement Table 5).

### 3.8 Analysis of co-expression genes associated with *CCL19* in breast cancer

The gene co-expression analysis module method of the Oncomine database was used to analyze the *CCL19* mRNA of TCGA breast cancer patients. As can be seen from Supplementary Fig. 8 A, the number of genes significantly positively

correlated with *CCL19* was higher than those of negative correlation. Supplementary Fig. 8B, C shows top 50 significant genes positively and negatively associated respectively. The results show that *CCL19* is highly correlated with SELP and KLRB1, reflecting that *CCL19* may take part in tumor development by mediating activated endothelial cells, inhibiting NK cells, and activating acid sphingomyelinase [22–24].

The TIBIS database provided us with the co-expression relationship of *CCL19* with immune activator, immunosuppressant, chemokine, chemokine receptor and MHC genes. The heatmap generated by the co-expression analysis indicated that a large number of immune-related genes were significantly positively correlated with the expression of *CCL19* in breast cancer (Supplementary Fig. 8D–H).

### 3.9 Correlation between *CCL19* expression and drug sensitivity in breast cancer

To explore the relationship between *CCL19* expression and common immune checkpoint-related genes, the expression of immune checkpoint-related genes in the tumor sample sequencing data was extracted for correlation testing. The correlation test filter was set to P-value < 0.001. The heat map showed the positive regulatory relationship between *CCL19* and multiple immune checkpoints, including LAIR1, ICOS, PDCD1 and TIGIT (Fig. 7A, B).

Integrate the immunotherapy effect scoring files in TCIA with the sorted TCGA sequencing data according to the sample ID. According to the expression of *CCL19*, the patients were divided into two groups: high and low expression. The “limma” package was used to compare the immunotherapy scores between the high and low expression groups. The “ggpubr” package was used to express the final difference comparison results into violin plots. The results of immunophenotypic analysis showed that the group with high *CCL19* expression exhibited higher IPS, indicating that patients with high *CCL19* expression exhibited greater positive responses to immunotherapy (Fig. 7C–F).

IC50 (semi-inhibitory concentration) is an important indicator for evaluating drug efficacy or sample treatment response and was used to analyze the sensitivity of patients with different *CCL19* expression levels to drugs. The results showed that patients with high *CCL19* expression had lower IC50 values for TAK-715 and KIN001-102 (Fig. 7G, H). In contrast, patients with high *CCL19* expression had higher IC50 values for 17-AAG (Fig. 7I).

### 3.10 Validation of the expression, prognosis, and immune-Related exploration of *CCL19* through clinical samples

IHC staining was performed on the cancer tissue and normal tissue samples from hospital patients. All staining steps were carried out in strict accordance with the standard procedure. The results showed that *CCL19* was darker brown in tumor tissue, which represented higher expression (Fig. 8A, B). For hospital patients, we compared the differential expression of *CCL19* in different groups. Breast cancer patients from hospitals were grouped based on age, nuclear grade, tumor stage, and molecular type (Table 1). Neither OS nor RFS based on clinical samples were found to be significantly associated (Supplementary Fig. 9). Thanks to early tumor stage and good treatment, breast cancer patients from our hospital have fewer recurrences and deaths within five years. In order to draw more accurate conclusions, it is necessary to further expand the research sample size in the future.

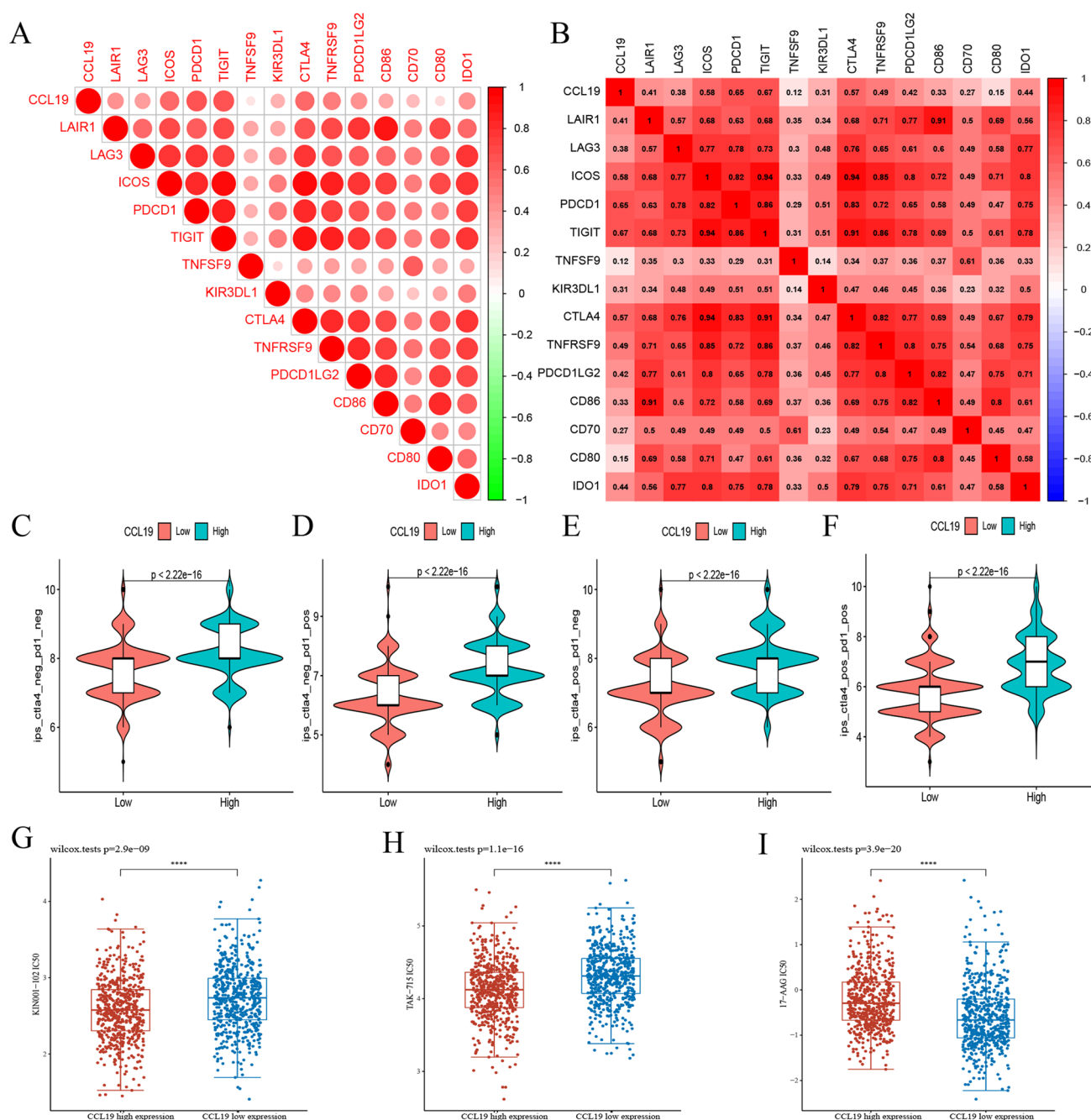
To explore the correlation between *CCL19* and immunotherapy, IHC expression levels of *CCL19* and CD8 were calculated. We found that *CCL19* expression was significantly correlated with the expression levels of CD8, PD-1 and PD-L1 (Fig. 8C).

*CCL11* inhibits the proliferation, migration, and invasion of BC cells.

Next, we analyzed the biological function of *CCL19* in breast cancer through cellular experiments. CCK-8 and colony formation experiments showed that the proliferation and colony formation of BC cells were inhibited in vitro after the addition of *CCL19* (Fig. 9A–E). To further evaluate the role of *CCL11* in BC cell progression, we investigated the migration and invasion of cells adding *CCL11* BC. Transwell experiment showed that the addition of *CCL11* significantly reduced the invasion and migration ability of BC cells (Fig. 9F–H).

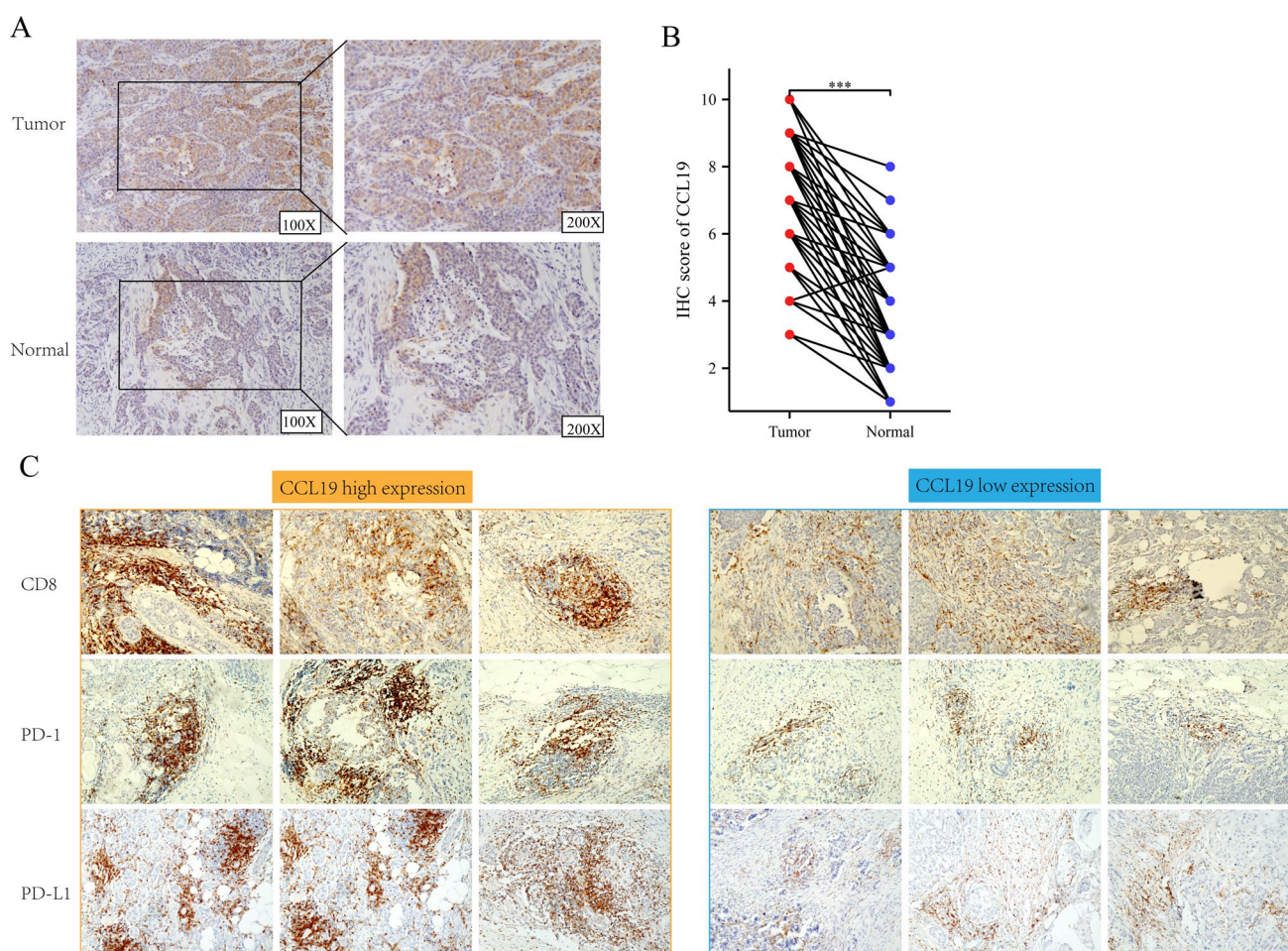
## 4 Discussion

In recent decades, significant improvements in breast cancer survival rates have been observed [25]. Nevertheless, the fundamental mechanisms driving breast cancer tumorigenesis remain only partially understood. Interactions between tumor cells and diverse components of the tumor microenvironment contribute to the overall complexity



**Fig. 7** **A, B** The association between ICIs and CCL19. **C–F** The correlation between immunophenoscore and different CCL19 expression groups. **G–I** IC50 value of KIN001-102, tak-715 and 17-AAG

of cancer [26]. Recent advancements in genomic research have significantly enhanced the understanding of breast cancer diagnosis and treatment. Genetic data, gene activation pathways, and gene-mediated functions are critical to understanding and treating cancer [27–29]. Tumors undergo transitions between elimination, equilibrium, and escape as the status of the immune system fluctuates [30, 31]. Hence, immune-related genes play a crucial role in tumor progression, garnering increasing attention in breast cancer diagnosis and treatment [32–34]. This study classified 21 breast cancer samples from the GSE108474 dataset into two groups based on immunotherapy responses, followed by differential expression gene (DEG) screening within the clusters. Then, DEG differentially expressed in the two clusters was screened. Key hub genes identified include *PTPRC*, *CST7*, *C1S*, *CD2*, *IL7R*, *CD3E*, *CD3D*, *GZMA*, *CCL5*,



**Fig. 8** Expression of *CCL19* in breast cancer patients from hospitals. **A** IHC images of *CCL19* in tumor and adjacent normal tissues; **B** IHC score of *CCL19* expression in tumor and adjacent tissues; **C** IHC images of CD8, PD-1 and PD-L1 expression in groups with *CCL19* high and low expression in breast cancer samples

*LYZ*, *CXCL9*, *NKG7*, *GBP5*, *CXCL11*, *CD79A*, *CXCL10*, *JCHAIN*, *CCL19*, *CCL7*, *CCL21*, and *CXCL13*. Considering the expression and prognostic implications of the central gene in BRCA, *CCL19* was identified as the key gene for this study.

The *CCL19* gene, situated at 9p13 on human chromosome 17, belongs to the CC chemokine family [3, 35]. *CCL19* is commonly viewed as a homeostatic chemokine that regulates immune cell trafficking, thereby contributing to cancer development. *CCL19* promotes lymphocyte homing and lymph node metastasis of tumor cells through its interaction with the CCR7 receptor [4]. *CCL19* exhibits a dual role in tumorigenesis. First, *CCL19* suppresses tumors by inducing anti-tumor immune responses and inhibiting angiogenesis. Zhuoqing Xu et al. concluded through cell experiments and animal experiments that *CCL19* can inhibit CRC angiogenesis by promoting the expression of miR-206 [5]. High expression of *CCL19* significantly inhibited the proliferation of GC cells and tumor growth in vitro and in vivo by up regulating the CCR7/AIM2 pathway [7]. Conversely, *CCL19* also plays a role in promoting tumor progression by inducing tumor cell proliferation and metastasis. Qian Liu et al. found that *CCL19* is linked to lymph node metastasis in small cell lung cancer patients and promotes tumor metastasis by impairing CD8 + T cell function [8, 34]. The knockout of *CCL19* attenuates tumor cell proliferation in cervical cancer [6].

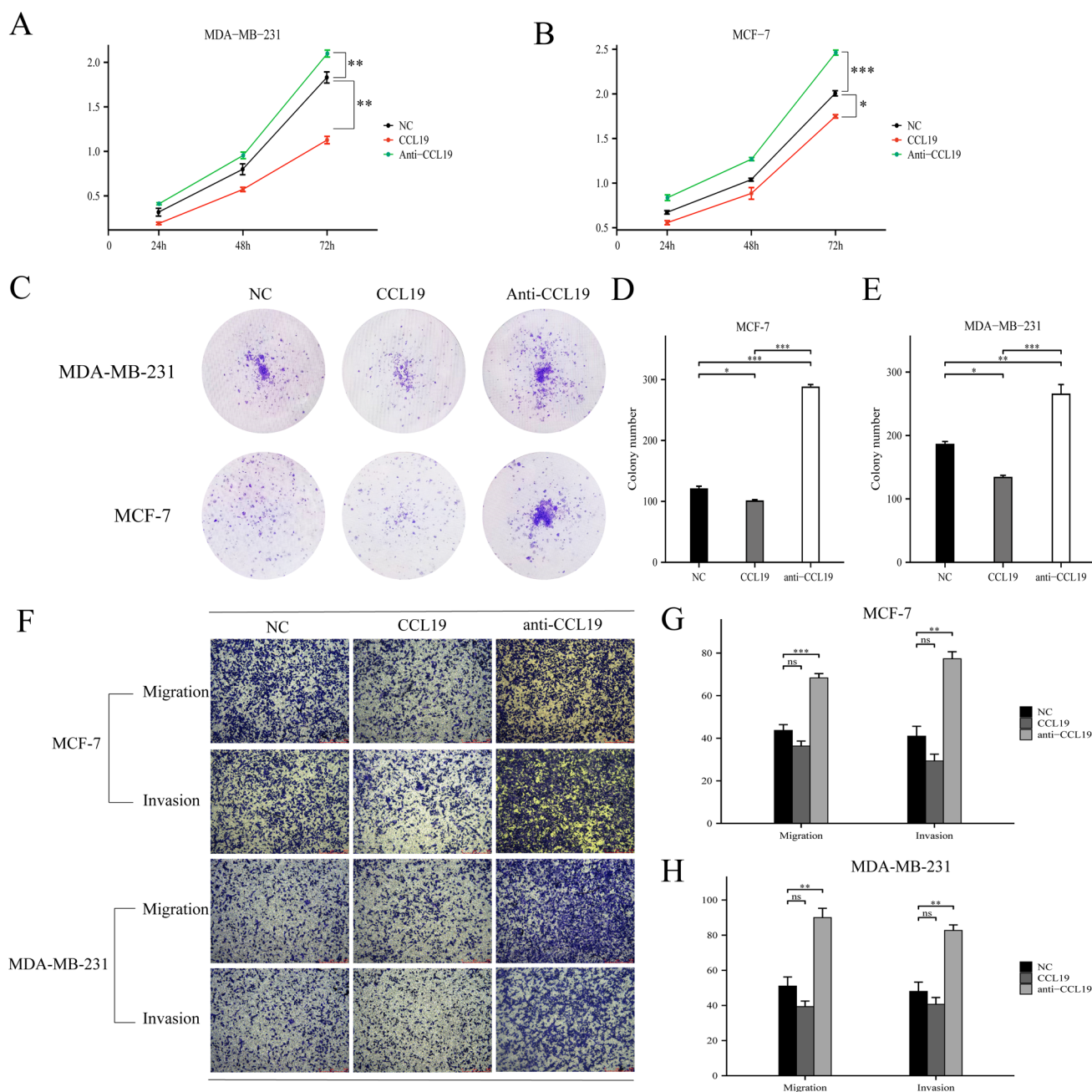
*CCL19* demonstrates both anti-tumor and pro-tumor activities in different contexts of breast cancer. *CCL19* has the potential to mediate strong anti-tumor responses in the body. The *CCL19*/CCR7 axis has been identified as regulating EMT and mediating tumor cell invasion and migration through AKT signaling pathway in breast cancer. Hossein Hozhabri et al. found that the overexpression of *CCL19* in breast cancer patients was associated with better OS and RFS [36]. *CCL19* can be combined and cleared by CX-CKR, which promotes breast cancer cell invasion and metastasis [37]. *CCL19* induces NK cells and inhibits tumor growth when breast cancer tumor-bearing mice are injected with *CCL19* [38, 39]. However, Xu et al. found that *CCL19* administration in vitro increased the invasion, migration and EMT of breast cancer cells. *CCL19*

**Table 1** Correlation between CCL19 expression and clinical characteristics of patients with BRCA (n = 101)

CLINICOPATHOLOGICAL Characteristics	Number	Low expression	High expression	P-value
Age (years)				0.456
≤ 55	48	20	28	
> 55	53	26	27	
Stage				0.036
Stage I	34	10	24	
Stage II	57	29	28	
Stage III	10	7	3	
T classification				0.327
T1	41	15	26	
T2	58	30	28	
T3	2	1	1	
N classification				0.053
N0	76	29	47	
N1	15	10	5	
N2	8	5	3	
N3	2	2	0	
Grade				0.034
1	18	9	9	
2	75	30	45	
3	8	7	1	
Subtype				0.506
Luminal A	18	9	9	
Luminal B	59	24	35	
HER-2 +	15	7	8	
Basal	9	6	3	
ER status				0.147
Positive	28	16	12	
Negative	73	30	43	
PR status				0.298
Positive	32	17	15	
Negative	69	29	40	
HER2 status				0.277
Positive	65	27	38	
Negative	36	19	17	
Ki-67 status				0.326
≤ 14	19	9	10	
> 15	82	37	45	

facilitates the secretion of IL-6 and IL-1 $\beta$  by DC cells, which promote tumor growth in triple-negative breast cancer (TNBC), leading to tumor invasion and worse prognosis in patients [40]. Recent research has identified *CCL19* as a potential prognostic biomarker that shows significant correlation with TILs in breast cancer [41]. Sangeetha Prabhakaran et al. identified 12 chemokine gene expression characteristics, including *CCL19*, capable of distinguishing breast cancer with heterotopic lymph node-like structures rich in tumor location, and predicting more favorable long-term outcomes [42].

This study revealed a significant upregulation of *CCL19* in breast cancer tumor tissue, which was positively correlated with patient prognosis. Data from the TCGA and GEPIA databases indicated a significant association between *CCL19* mRNA expression levels, cancer stage, and tumor subtype in breast cancer patients. IHC results from hospital patients further corroborated this finding. In addition, the mutation rate of *CCL19* in Invasive ductal carcinoma patients was 1.39%. Amplification represents the most common type of *CCL19* mutation in breast cancer. *CCL19* expression exhibited a negative correlation with MSI and TMB in breast cancer. The current study identified a



**Fig. 9** CCL19 inhibits malignant behavior of breast cancer cell lines. **A, B** After adding CCL19, the proliferation capacity of breast cancer cells was measured by CCK-8; **C–E** After adding CCL19, the proliferation and viability of breast cancer cells were detected by plate cloning; **F–H** Transwell assay evaluated the migration and invasion ability of breast cancer cells after adding CCL19

significant correlation between *CCL19* expression and the infiltration levels of various TILs in the tumor microenvironment. These findings support the hypothesis that *CCL19* regulates tumor progression by modulating the trafficking of immune cells, including DCs, T cells and NK cells. Research has demonstrated that *CCL19* induces maturation and migration of DCs while reducing apoptosis [43]. *CCL19* indirectly increases T cell proliferation by promoting the production of IL-12, TNF $\alpha$  and IL-1 $\beta$  by DC cells (20). *CCL19* can increase the proliferation of NK cells triggered by IL-2 (20,21). *CCL19* augments T cell responses by recruiting and facilitating their steady-state transport along a concentration gradient of *CCL19* [44]. This study identified a positive correlation between *CCL19* and the density of infiltrating CD8 + T cells in the TME within our cohort, suggesting a potential role of *CCL19* in regulating breast cancer progression through CD8 + T cells. Functional network analysis suggested that *CCL19* may contribute to breast cancer

progression by engaging in signaling pathways such as the PI3 K-Akt pathway and Human Papillomavirus infection. Peptides derived from Human Endogenous Retroviruses (HERV) have been shown to stimulate *CCL19* production in pancreatic cancer cells via the PI3 K pathway [45]. *CCL19* enhances Akt phosphorylation and upregulates Bcl-2 protein expression in HNSCC cells. Elevated Bcl-2 expression inhibits cisplatin-induced apoptosis, implying a significant role of *CCL19* in chemoresistance [46].

Chemotherapy is widely recognized as a potent inducer of apoptosis. Tumor cells evade apoptosis by down-regulating pro-apoptotic pathways, upregulating anti-apoptotic signals, or inducing false apoptotic signals [47]. Hence, evaluating the sensitivity of breast cancer patients with varying *CCL19* expression levels to chemotherapy is imperative. The results demonstrated a significant correlation between *CCL19* expression levels and the IC50 values of TAK-715, KIN001-102, and 17-AAG. Research has demonstrated that direct intratumoral injection of *CCL19* significantly reduces tumor volume in a mouse model of lung cancer [10, 48]. Xu et al. observed significant inhibition of intratumoral angiogenesis following treatment with *CCL19* in both colon cancer cells and mice [5]. Co-injection of *CCL19* and cytosine-phosphorothioquinine oligonucleotides (CpG ODN) into tumor sites in a mouse model led to substantial tumor shrinkage [49]. Treatment with *CCL19* markedly upregulated Bcl-xL and Bcl-2 proteins in breast cancer cells, thereby countering anoikis [50]. The combination of *CCL19*-encoding plasmid DNA (*CCL19* pDNA) and PD-L1 inhibitor (BMS-1) significantly inhibited tumor development..

In this study, we employed multiple bioinformatics approaches to screen for hub genes in BRCA samples, identifying *CCL19* as the most prominent gene. Although our findings are promising, several limitations remain in this study. First, the survival analysis results are limited by the generally favorable prognosis associated with breast cancer. In our cohort, no significant differences in OS and RFS were observed among groups with varying levels of *CCL19* expression. A longer follow-up period is required to thoroughly assess the impact of *CCL19* on patient prognosis. Second, the precise mechanisms and pathways through which *CCL19* influences breast cancer progression remain unclear. Finally, while we evaluated the drug sensitivity of *CCL19* using the GDSC database, further experimental validation is necessary. Additional experiments are essential for further verification. Addressing these limitations will guide future research directions.

In conclusion, our study identifies *CCL19* as a promising biomarker for breast cancer. However, the mechanisms underlying *CCL19*'s role in regulating tumor progression in breast cancer remain poorly understood. Thus, future research will focus on elucidating the genetic axis of *CCL19* in breast cancer.

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**Author contributions** XC contributed to conception and design of the study. YW organized the database. LY and XW performed the statistical analysis. XC and ZD wrote the first draft of the manuscript. All authors contributed to manuscript revision, read, and approved the submitted version.

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**Data availability** Data is provided within the manuscript or supplementary information files. The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.

## Declarations

**Ethics approval and consent to participate** The study was approved by the Institutional Research Ethics Committee of The Second Affiliated Hospital of Anhui Medical University (Hefei, China), and written informed consent was obtained from all patients.

**The accordance statements** All experimental methods in this study were conducted in accordance with relevant guidelines and regulations.

**Consent for publication** Not applicable.

**Competing interests** The authors declare no competing interests.

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