KIF2C is a Biomarker Correlated With Prognosis and Immunosuppressive Microenvironment in Human Tumors

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Kinesin superfamily member 2C (KIF2C) is an essential regulator of the cell cycle and its aberrant expression can promote tumor progression. However, the mechanism of KIF2C in pan-cancer is unclear. Data were obtained from public databases, including The Cancer Genome Atlas (TCGA), UALCAN, TIMER, and CellMiner. We analyzed the correlation of KIF2C with expression, prognosis, tumor mutation burden (TMB), microsatellite instability (MSI), mismatch repairs (MMR), immune infiltration and anticancer drug sensitivity by R language. KIF2C was highly expressed in several tumors and correlated with poor prognosis. KIF2C expression was significantly correlated with TMB, MSI, MMRs, and immune checkpoint genes, and with the level of immune cell infiltration such as tumor-associated macrophage (TAM), cancer-associated fibroblasts (CAFs), myeloid-derived suppressor cells (MDSCs) and Tregs. The GO and KEGG results suggest that KIF2C is involved in immune regulation in addition to cell cycle regulation. In addition, KIF2C is associated with DNA methylation, m6A modifications and m7G modifications. Our data suggest that KIF2C is a prognostic biomarker linked to immunosuppression, targeting KIF2C may improve the outcome of immunotherapy. Our findings indicate that KIF2C is a prognostic biomarker associated with immunosuppression, and that targeting KIF2C may improve the outcome of immunotherapy.

Keywords: KIF2C, pan-cancer, tumor microenvironment, immunotherapy, biomarker

INTRODUCTION

The Kinesin-13 Family, which includes KIF2A, KIF2B, KIF2C, and KIF24, is engaged in mitotic regulation (Lucanus and Yip, 2018). KIF2C (Kinesis superfamily member 2C) is the most characteristic member of the Kinesin-13 family (Rath and Kozielski, 2012). By regulating microtubule dynamics, KIF2C performs many functions in spindle assembly, sister chromatid separation, and error correction. However, in tumor cells, the regulation of KIF2C is disturbed and instead enhances mitotic defects and promotes tumor progression (Ritter et al., 2016). Previous studies have shown that KIF2C is overexpressed in some tumors and promotes their progression (Eichenlaub-Ritter, 2015). In cervical cancer (CC), KIF2C promotes its proliferation by inhibiting the P53 pathway (Yang et al., 2022). KIF2C is upregulated in endometrial cancer (EC), which can cause a malignant phenotype in EC and negatively correlates with CD8 T cells, which have cancer suppressive effects (An et al., 2021). In hepatocellular carcinoma (HCC), KIF2C can promote...
Finally, the "immune cell in" the of KIF2C in tumor and normal tissues was performed by the different tumor types. In this study, the differential expression database that allows the analysis of immune infiltration, and thus promote tumor progression. (Petitprez et al., 2020). Soluble factors secreted by tumor or stromal cells can also induce the development of immune resistance, such as TGFβ, FGF2, and PDGF secreted by CAFs (Sadeghi Rad et al., 2021). Therefore, it is crucial to explore a new marker to improve the effectiveness of immunotherapy.

In this work, we investigated the prognosis of KIF2C in pan-cancer and its relationship with immune infiltration, and found that KIF2C was associated with cell cycle after enrichment analysis, and finally, we investigated the correlation between KIF2C expression and anticancer drug sensitivity. Our study demonstrates that KIF2C can serve as a prognostic marker in several tumors. They are correlated with immune infiltration and may be promising targets for enhancing the curative effect of immunotherapy.

MATERIALS AND METHODS

The Cancer Genome Atlas

TCGA is a public platform containing oncogene information. The expression data, TMB data, MSI data and clinical data of TCGA for 33 tumor types can be downloaded through UCSC Xena online database (https://xenabrowser.net/) (Goldman et al., 2020). The abbreviations and meanings of the 33 tumor types are given in the Supplementary Table S1.

UALCAN

UALCAN (http://ualcan.path.uab.edu) (Chen et al., 2019) is an online database that allows the analysis of gene expression data from TCGA. In the current study, we evaluated protein expression levels of KIF2C in LIHC using the "CPTAC analysis" module.

Tumor Immune Estimation Resource 2.0

TIMER 2.0 (http://timer.cistrome.org/) (Li et al., 2017) is a database that allows the analysis of immune infiltration in different tumor types. In this study, the differential expression of KIF2C in tumor and normal tissues was performed by the "Gene_DE" module, the correlation of KIF2C expression with the level of immune infiltration in 33 tumor types was visualized by the "Gene" module, and the prognosis of KIF2C expression and immune cell infiltration was analyzed by the "Outcome" module. Finally, the "Correlation" module was used to analyze the relationship between KIF2C expression and immune marker subgroups in KIRC, LGG, LIHC and TYHM.

Gene Enrichment Analysis

GO and KEGG enrichment analysis can be used as a way to explore gene function. In this study, we divided the expression of KIF2C into high and low groups according to the median and then performed enrichment analysis using clusterprofiler in R.

Drug Sensitivity Analysis

CellMiner (Shankavaram et al., 2009; Reinhold et al., 2012) is a database that integrates molecular and pharmacological data from NCI-60 cancer cell lines. Data on drug sensitivity was obtained from CellMiner, p <0.05 was used as a screening condition and correlation analysis was performed by the spearman’s test. Computational Analysis of Resistance (CARE) (jiang et al., 2018) is a database that can be used to calculate CARE scores to predict the correlation between target genes and the efficacy of targeted drugs.

GeneMania

GeneMania is an online database that can predict the function of specific genes by building protein-protein interaction networks (Warde-Farley et al., 2010). The selected genes can be analyzed by bioinformatics methods. In this study, we constructed a PPI network for KIF2C through GeneMania to explore the function of KIF2C. In this study, we constructed a PPI network of KIF2C through GeneMania to explore the function of KIF2C.

Immunohistochemistry (IHC)

To explore the protein expression level of KIF2C, we examined the expression level of KIF2C in paraffin-embedded tissues of liver cancer and normal liver tissues by IHC staining. Tissue sections were treated with anti-KIF2C primary antibody (1:200, Proteintech), and the protocol is described in (Zhang et al., 2021).

Cell Culture and Quantitative Real-Time PCR

All cell lines in this study were obtained from the Oncology Laboratory of Tongji Hospital, Huazhong University of Science and Technology. Among them, BGC-823, SGC7901, and LO2 cell lines were cultured by RPMI-1640 complete medium, while MGC-803, Hep-3B, and MHCC97-H cell lines were cultured by DMEM complete medium at 37°C in 5% CO2.

Total RNA was extracted from various cell lines using TRIzol reagent (Takara, Shiga, Japan). The mRNA expression levels of KIF2C in each cell line were subsequently verified by RT-PCR according to the SYBR-Green (Takara, Shiga, Japan) ’s protocol. The relative expression was calculated by 2-ΔΔC t and normalized using GAPDH as an internal reference.

Statistical Analysis

We performed statistical analysis via R (version 4.1.0), forestplot based on the forestplot package for visualizing the hazard ratio (HR) of KIF2C in pan-cancer, survival analysis via the survminer
and survivor packages, box plots and scatter plots via the ggplot2 package, and reshape2 package to plot correlation heat map. ROC curves are plotted via the ROC package, and radar plots are visualized via the ggradar package. The correlation of KIF2C expression with TMB, MSI, MRR immune checkpoint genes and drug sensitivity was analyzed by the spearman test, and the analysis of the difference in pan-cancer expression was analyzed by Wilcox test. \( p < 0.05 \) was accepted as statistically significant.

RESULTS

KIF2C is Overexpressed in Multiple Tumors

We analyzed the mRNA levels of KIF2C in 33 tumor types by TIMER 2.0 database and GEPIA database and found KIF2C was highly expressed in BLCA, BRCA, CESC, CHOL, COAD, DLBC, ESCA, GBM, HNSC, KIRC, KIRP, LGG, LIHC, LUAD, LUSC, OV, PCPG, PRAD, READ, SKCM, STAD, THYM, UCEC, and UCS (Figures 1A,B). Moreover, the protein level of KIF2C was analyzed by UALCAN database and found that KIF2C was significantly overexpressed in BRCA, KIRC, COAD, OV, LUAD, and LIHC (Figure 1C). In addition, we verified the mRNA expression levels of KIF2C, and we found that KIF2C expression was significantly higher in human HCC cell lines (MHCC97-H, Hep-3B) than in normal human hepatocytes (LO2) (Supplementary Figure S1A). However, there was no significant difference in the expression of KIF2C in gastric cancer cell lines (Supplementary Figure S1B). We analyzed the mRNA levels of KIF2C in 33 tumor types by the TCGA database and GTEx database, and combined with differential analysis of paired samples, we found that KIF2C was highly expressed in ACC, BLCA, BRCA, CHOL, COAD, DLBC, ESCA, HNSC, KIRC, KIRP, LGG, LIHC, LUAD, LUSC, OV, PCPG, PRAD, READ,
SKCM, STAD, THYM, UCEC, and UCS (Figures 1A,B). In addition, analysis of KIF2C protein levels by the UALCAN database revealed that KIF2C was overexpressed in LIHC (Figure 1C). We verified the expression level of KIF2C in liver cancer tissues by immunohistochemistry, and the results were consistent with the database (Figures 1D,E). In addition, we verified the mRNA expression levels of KIF2C, and we found that KIF2C expression was significantly higher in human HCC cell lines (MHCC97-H, Hep-3B) than in normal human hepatocyte lines (LO2) (Supplementary Figure S1A). However, there was no significant difference in the expression of KIF2C in gastric cancer cell lines (Supplementary Figure S1B).

**KIF2C is a Potential Prognostic Marker for Multiple Tumors**

We divided the tumors into two groups, high and low expression, according to the median expression of KIF2C in various tumor types. We investigated the prognostic value of KIF2C in TCGA pan-cancer by univariate Cox regression analysis. The results are presented in the form of a forest plot (Figure 2A). KIF2C overexpression could significantly impact ACC (HR = 2.398), KICH (HR = 2.708), KIRC (HR = 2.142), KIRP (HR = 2.768), LGG (HR = 1.692), LIHC (HR = 1.586), LUAD (HR = 1.210), MESO (HR = 2.137), PAAD (HR = 1.577), PRAD (HR = 2.290), SARC (HR = 1.262), THYM (HR = 0.538), and UCEC (HR = 1.497) for overall survival. KIF2C upregulation related to poorer OS prognosis of the above tumors, except for TYHM. Except for TYHM, KIFU2C upregulation is associated with a poorer OS prognosis.

Subsequently, we performed Kaplan-Meier survival analysis and found that KIF2C upregulation was significantly associated with poorer prognosis in ACC, KIRC, KIRP, LGG, LIHC, MESO, and UCEC, while it was significantly associated with better prognosis in THYM (Figure 2B).

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**FIGURE 2** OS analysis of KIF2C in pan-cancer (A) Forest plot shows hazard ratios of KIF2C in pan-cancer. (B–I) Kaplan-Meier survival curves of pan-cancer with high and low KIF2C expression analysed by the TCGA database.
In addition, we also investigated the correlation of KIF2C with PFI, DFI and DSS, and the results showed that KIF2C was significantly associated with poor prognosis in several cancer types (Supplementary Figure S2). The results of OS, PFI, DFI and DSS survival analysis consistently indicated that KIF2C was a risk factor for ACC, KIRP, LIHC, LUAD, PRAD, SARC, and UCEC.

Correlation of KIF2C Expression With TMB, MSI and DNA Mismatch Repair (MMR) in Multiple Tumors

Previous studies have shown that TMB and MSI-H are predictive biomarkers for tumor immunotherapy (Rizzo et al., 2021), thus we analyzed the correlation between KIF2C expression and different tumor TMB, MSI and MMRs. Figure 3A shows that high KIF2C expression was significantly correlated with TMB of 24 tumors, and only significantly negatively correlated with THYM. Figure 3B shows that high expression of KIF2C was positively correlated with MSI in BLCA, COAD, HNSC, LIHC, LUSC, SARC, STAD, UCEC, and UCS. dMMR can induce MSI production, so we evaluated the association between the expression level of KIF2C and the mutation levels of five MMR genes (MLH1, MSH2, MSH6, PMS2, and EPCAM). Figure 3C shows that KIF2C was significantly associated with MMR genes in 30 tumors.
Epigenetic Analysis of KIF2C
We performed methylation analysis of KIF2C through the GSCA database (Liu et al., 2018) and found that KIF2C was significantly negatively correlated with KIF2C methylation levels in most tumors (Figure 4A), and KIF2C hypomethylation was associated with poor prognosis of LGG and CHOL by prognostic analysis (Figure 4B). We further investigated the correlation between KIF2C and DNA methyltransferase (DNMT)-related genes, N6-methyladenosine (m6A) and 7-methylguanosine (m7G)-related genes, and the results showed that KIF2C was significantly correlated with the above-mentioned related genes in a variety of tumors (Figure 4C), suggesting that KIF2C is closely related to epigenetic modifications that affect tumor progression.

Correlation Between KIF2C Expression and Tumor Microenvironment of Multiple Tumors
TME plays an important role in promoting tumor progression and consists mainly of immune cells and stromal cells (Binnewies et al., 2018). We obtained immune scores and stromal scores of 33 tumors from TCGA database by the ESTIMATE algorithm and performed correlation analysis. Figure 5 showed that KIF2C expression was significantly positively correlated with immune and stromal scores in KIRC, LGG, and THCA, but negatively correlated with those in GBM, LUSC, STAD, and UCEC. In addition, high KIF2C expression was negatively correlated with stromal scores in BRCA, COAD, HNSC, LIHC, LUAD, SARC, SKCM, TGCT, and THYM, and negatively correlated with immune scores in DLBC (Supplementary Figure S3).

Immune Infiltration Analysis of KIF2C
Using the TIMER database, we examined the correlation between KIF2C expression and immune infiltration levels in various cancers. In Figure 6A, KIF2C expression in KIRC, LGG, LIHC, and TYHM was significantly correlated with B cell, CD8+ T cell, CD4+ T cell, macrophage cell, neutrophil cell, and dendritic cell. Except for the negative correlation between KIF2C and the neutrophil cell of THYM, all of them were significantly positively correlated. By KM analysis (Figure 6B), high KIF2C expression and high infiltration levels of the six immune cells were significantly correlated with poor OS in LGG, whereas high KIF2C expression and high infiltration levels of B cells, CD4+ T cells, and dendritic cells were correlated with better OS in THYM.

Myeloid-derived suppressor cells (MDSCs) have immunosuppressive characteristics, and we analyzed the correlation between KIF2C and MDSCs in pan-cancer by TIMER 2.0 and found that KIF2C was significantly positively correlated with MDSCs in most tumors (Figure 7A). Moreover, MDSCs was associated with a poorer prognosis (Figure 7B).

Additionally, we used EPIC, CIBERSORT-ABS, and QUANTISEQ algorithms to analyze the correlation of KIF2C with cancer-associated fibroblasts, M2 macrophages, and Tregs in KIRC, LGG, LIHC, and THYM. We found that KIF2C was positively correlated with CAFS, M2 macrophages, and Tregs in KIRC, LGG, and LIHC. However, in THYM, KIF2C was negatively correlated with all three of these cells (Figure 8). We then analyzed the association between KIF2C expression and immune markers in KIRC, LGG, LIHC, and TYHM, and to reduce bias, we corrected the data according to tumor purity. Supplementary Table S2 shows that KIF2C expression in KIRC was mainly positively correlated with...
TAM, M2 macrophages, Tregs, and T cell exhaustion. In contrast, in TYHM, KIF2C expression was negatively correlated mainly with M2 macrophages, monocytes, T cell exhaustion, and TAM. It is evident that KIF2C has an important role in immune infiltration.

**Correlation of KIF2C Expression With immune Checkpoint Genes and immune Response in Multiple Tumors**

To explore the relationship between KIF2C and immunotherapy, we performed a pan-cancer correlation analysis of KIF2C with 47 immune checkpoint (ICP) genes. Supplementary Figure S4 shows that KIF2C expression correlated with 42 genes in THCA, 39 genes in PRAD and LIHC, and 38 genes in BRCA and THYM. More than 30 ICP genes were associated with KIF2C in BLCA, BRAC, HNSC, KICH, KIRC, KIRP, LGG, LIHC, MESO, PAAD, and THCA, suggesting that KIF2C may exert an immunosuppressive effect in these tumors and that targeting KIF2C can achieve better immunotherapy outcomes. However, KIF2C is mainly negatively correlated with ICP genes in COAD, DLBC, GBM, LUAD, LUSC, READ, SKCM, STAD, THYM, and UCEC, which means that
targeting KIF2C in such tumors may not achieve favorable immunotherapy outcomes.

To explore the KIF2C expression effect on immune checkpoint blockade (ICB) treatment response more deeply, we used the TIDE algorithm to obtain the TIDE scores of KIF2C in KIRC, LGG, LIHC and THYM. High TIDE scores were associated with poor response to ICB therapy and short survival after receiving ICB therapy (Jiang et al., 2018). We found that the KIF2C high expression group had higher TIDE scores in KIRC, LGG and LIHC, while it had lower TIDE scores in THYM (Figures 9A–D), suggesting that KIF2C may promote immune escape and cause poor ICB response in KIRC, LGG and LIHC, but in THYM the opposite is true. This is consistent with our previous findings.

**GO and KEGG Analysis of KIF2C Expression**

To explore the biological role of KIF2C in tumors, we performed GO and KEGG analysis on three tumors, KIRC, LIHC, and THYM. As shown in Figure 10, we found that KIF2C is involved in the regulation of the cell cycle in all three tumors and is related to immune regulation. In KIRC, KIF2C is involved in immune-related processes such as “complement and coagulation cascade,” “cytokine-cytokine receptor interaction,” and also, KIF2C is associated with “Coronavirus disease-COVID-19” in LIHC. KIF2C was correlated with “T cell receptor signaling pathway” and “T cell activation” in THYM. It can be seen that KIF2C is closely related to immunity but plays a different immune role depending on the type of cancer.

![Figure 6](image-url)
PPI Network of KIF2C Analysis
We constructed the PPI network of KIF2C through the GeneMania database. Figure 11 shows that KIF2C interacts with genes such as CENPH, ADAM9, NDEL1, and EXT1 and is associated with biological functions such as tubulin binding, sister chromatid segregation, and MHC II antigen presentation, suggesting that KIF2C is involved in cell cycle and immune regulation. In addition, previous studies found that the KIF2C-related CENPH, EXT1, can promote tumor progression (Wu et al., 2015; Lu et al., 2017; Kong et al., 2021) and ADAM9 has a significant role in tumor growth, metastasis, and immune evasion (Chou et al., 2020).

Drug Sensitivity Analysis of KIF2C
Using the CellMiner database, we looked at the relationship between KIF2C expression and drug sensitivity. As shown in Figure 12A, KIF2C expression was significantly positively related to the sensitivity of Megestrol acetate, Rebimastat, and Ibrutinib. In contrast, the sensitivity of Tamoxifen, Denileukin Diftitox Ontak and Fluorouracil was significantly negatively correlated. To further investigate, we analyzed the correlation between KIF2C expression and drug efficacy through the CARE database. The results showed that KIF2C expression was positively correlated with the sensitivity of PI3K-AKT signaling pathway inhibitors (MK-2206, Rigosertib), CDK1 inhibitor (Flavopiridol) and PLK1 inhibitor (BI2536), while it was correlated with the resistance to EGFR inhibitors (Figure 12B). Detailed information is presented in Supplementary Table S3.

Construction of Nomogram
We verified the predictive efficacy of KIF2C in tumors by ROC predictive effect on the prognosis of the above cancer types. In order to predict the prognosis of patients more accurately, we constructed a nomogram based on KIF2C for LGG and LIHC to predict the 1,3,5-year survival rate of patients and validated it with the calibration curves (Figures 13E–H). We found that the nomogram was in good agreement with the actual situation.

DISCUSSION
The function of KIF2C in cell division requires complex regulation by various kinases and phosphatases, such as Aurora A/B, Plk1, and Cdk1 (Sanhaji et al., 2011). However, when KIF2C is deregulated, it causes abnormal spindle formation and failed chromosome segregation, resulting in chromosomal instability and tumorigenesis (Sanhaji et al., 2011). Previous studies have identified that KIF2C was overexpressed in some cancers (Nakamura et al., 2007; Shimo et al., 2007; Ishikawa et al., 2008; Wei et al., 2020) and facilitates the growth and invasion of tumor cells. Currently, the function of KIF2C in tumors is not clear, and a comprehensive pan-cancer analysis is needed.

In this study, the mRNA levels of KIF2C were highly expressed in 28 tumors by TCGA database analysis. In addition, analysis of the UALCAN database revealed that the protein levels of KIF2C were significantly higher in HCC tumors than in adjacent tissues. The prognostic analysis revealed that KIF2C was negatively correlated with the OS of ACC, KIRC, KIRP, LGG, LIHC, MESO, and UCEC, and positively correlated with the OS of THYM only, which, combined with the results of Cox regression analysis, can be considered as a prognostic marker in several cancers.

Epigenetic alterations have important effects on tumor progression and immunosuppression, of which DNA methylation is one of the most important epigenetic regulators (Dawson and Kouzarides, 2012; Jones et al., 2019). We found that the mRNA levels of KIF2C were significantly correlated with methylation in different cancer types, and there was a correlation between DNMT and KIF2C expression, suggesting that DNA methylation may also be involved in the regulation of KIF2C. In LGG and CHOL, hypomethylation of KIF2C was significantly associated with poor prognosis. The mechanism may be related to the hypomethylation of KIF2C leading to tumorigenesis and overexpression of immunosuppression-related genes.

In addition to DNA methylation, internal modifications of RNA are also important and play an important role in tumor
progression, especially m6A modification and m7G modification (Barbieri and Kouzarides, 2020). m6A modification can affect immune cell infiltration in the tumor microenvironment (Zhang et al., 2020), while m7G modification can cause immune escape (Devarkar et al., 2016). Therefore, we also explored the correlation between KIF2C and m6A and m7G-related genes, and found that KIF2C was significantly associated with m6A and m7G-related genes in most tumors, suggesting that the above RNA modifications may be involved in the regulation of KIF2C and thus influence the tumor progression.

Immunotherapy is emerging as an important modality in the treatment of cancer, such as ICB targeting CTLA-4 and PD-1 and its ligand (PD-L1), which can be used to treat tumors by reactivating anti-cancer immune responses (Petitprez et al., 2020). However, it was found that the
interaction between the tumor cells in the TME and their surrounding stroma could affect the therapeutic effect of ICB and even cause drug resistance. Hence, the correlation between KIF2C and TME was explored. Analysis by the ESTIMATE algorithm revealed that KIF2C expression was significantly positively correlated with immune and stromal scores in KIRC, LGG, and THCA, and significantly negatively correlated in GBM, LUSC, STAD, and UCEC, suggesting a possible role of KIF2C expression in tumor immunity and TME. We analyzed the relationship between KIF2C expression and immune cell infiltration through the TIMER database and found that KIF2C expression was significantly correlated with all six immune cell types in KIRC, LGG, LIHC, and THYM, and by correlation with immune markers, we discovered that KIF2C expression was positively correlated with CAFs, TAMs, M2 macrophages, Tregs, and MDSCs. TAMs are usually dominated by the M2 subtype, which inhibits tumor progression, the M2 subtype can promote tumor cell invasion, tumor cell invasion and metastasis, and the generation of immune tolerance (Shu and Cheng, 2020). whereas Tregs can lead to T cell exhaustion and are contributes to tumor evasion from immune surveillance and is a major obstacle to immunotherapy. While Tregs can cause T cell exhaustion and contribute to tumor evasion from immune surveillance, which is a major barrier to immunotherapy (Wolf et al., 2015), MDSCs is a group of myeloid cells that arise under pathological conditions of cancer, MDSCs can promote tumor progression as well as tumor immunosuppression. MDSCs are a group of myeloid cells that arise under pathological conditions of cancer and can promote tumor progression as well as tumor immunosuppression (Gabrilovich, 2017), CAFs are important components of TME and can promote tumor invasion, metastasis, and chemoresistance. In addition, CAFs can promote the
recruitment, activation, and immunosuppressive behavior of immunosuppressive cells, including M2-type TAMs, Tregs and MDSCs (Mao et al., 2021). The above indicates that KIF2C expression is associated with immunosuppression.

In THYM, KIF2C strongly correlated with CD8+ T cells with tumor suppressive effects and negatively correlated with CAFs, M2 macrophages, Tregs, and TIDE score, suggesting that KIF2C expression in THYM is a protective factor that predicts a better ICB response. In contrast, in KIRC, LGG, and LIHC, KIF2C was positively correlated with CAFs, M2 macrophages, Tregs, MDSCs, and TIDE scores, suggesting that KIF2C high expression in these three tumors is associated with immunosuppression and poorer immunotherapeutic response, and thus KIF2C is a prognostic risk factor for KIRC, LGG, and LIHC, and targeting KIF2C may improve immunotherapeutic outcomes.

Previous studies have shown that defective MMR leads to MSI-H (Lu et al., 2021), and MSI and TMB are considered to be predictable markers predictors of ICB response (Chang et al., 2018; Sha et al., 2020), so the correlation of KIF2C with MMRs, MSI, and TMB was explored and it was found that KIF2C expression was significantly correlated with TMB in 24 tumors, MSI in 9 tumors, and MMRs in 30 tumors. We also analyzed the relationship between KIF2C and 47 ICP genes and found that KIF2C expression was significantly associated with several ICP genes, suggesting that KIF2C can be considered as a biomarker to predicting predict the response of to immunotherapy. We analyzed the biological functions of KIF2C by gene enrichment analysis. We found that in addition to the cell cycle, KIF2C expression was also significantly associated with complement and coagulation cascade, IL-17 signaling pathway, and cytokine-cytokine

FIGURE 10 | Gene enrichment analysis of KIF2C in KIRC, LGG, and THYM. (A–C) Top 20 GO and KEGG pathway analysis in the specified tumor types.
FIGURE 11 | PPI network of KIF2C constructed using the GeneMania database.

FIGURE 12 | Anticancer drug sensitivity analysis of KIF2C. (A) CellMiner’s drug sensitivity analysis. (B) CARE analysis
receptor interaction, suggesting that KIF2C is involved in tumor regulation of immunity.

We found that EXT1 was associated with KIF2C by PPI network analysis. And EXT1 may affect tumor survival by activating the Wnt signaling pathway. By PPI network analysis, we found that EXT1 was associated with KIF2C. EXT1 may affect tumor survival by activating the Wnt signaling pathway (Kong et al., 2021), and it has also been demonstrated that KIF2C is activated by the Wnt signaling pathway and thus can promote liver cancer progression by activating mTORC1 signaling (Wei et al., 2020), so we speculate that EXT1 may influence KIF2C and thus promote tumor development.

Finally, we investigated the connection between KIF2C and anticancer drug resistance using CellMiner and found that KIF2C was negatively correlated with the sensitivity of Tamoxifen, Denileukin Diftitox Ontak, and Fluorouracil. Tamoxifen can be used to treat breast cancer (Shagufta and Ahmad, 2018), while Fluorouracil is one of the main components of many chemotherapy regimens and plays an important role in tumor chemotherapy (Ghafoori-Fard et al., 2021). High KIF2C expression was found to be associated with resistance to these drugs, and therefore, targeting KIF2C may have a role in improving the efficacy of chemotherapy treatment.

In summary, KIF2C was highly expressed in multiple tumor tissues and was significantly associated with poor prognosis. In tumor immunity, KIF2C was significantly associated with CAFs, TAMs, Tregs, and MDSCs infiltrates in TME. KIF2C was significantly correlated with TMB, MSI, MRR, and ICP genes in several tumors, and KIF2C expression was correlated with the sensitivity of some anticancer drugs. In tumor immunity, KIF2C was significantly associated with CAFs, TAMs, Tregs, and MDSCs infiltrating the TME. In several tumors, KIF2C expression was significantly correlated with TMB, MSI, MRR, and ICP genes, and KIF2C expression was correlated with the sensitivity of some anticancer drugs. Therefore, KIF2C is a prognostic biomarker linked to immunosuppression, targeting KIF2C may improve the outcome of immunotherapy.
DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

AUTHOR CONTRIBUTIONS

XZ contributed to data acquisition and article drafting; YL, PH, and LX provided technical support; HQ contributed to study design and supervision. All authors read and approved the final article.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fgene.2022.891408/full#supplementary-material
Shimo, A., Tanikawa, C., Nishidate, T., Lin, M.-L., Matsuda, K., Park, J.-H., et al. (2007). Involvement of Kinesin Family Member 2C/mitotic Centromere-Associated Kinesin Overexpression in Mammary Carcinogenesis. Cancer Sci. 99, 62–70. doi:10.1111/j.1349-7006.2007.00635.x

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