Check for updates

OPEN ACCESS

EDITED BY Rui Cao, Affiliated Beijing Friendship Hospital, Capital Medical University, China

REVIEWED BY Fahui Liu, University of Gothenburg, Sweden Shenglin Gao, Changzhou No.2 People's Hospital, China

*CORRESPONDENCE

Junbin Yuan, i dabing7220@126.com Longxiang Wu, i wulongxiang1123@126.com

[†]These authors have contributed equally to this work

SPECIALTY SECTION

This article was submitted to Cancer Genetics and Oncogenomics, a section of the journal Frontiers in Genetics

RECEIVED 16 November 2022 ACCEPTED 08 December 2022 PUBLISHED 04 January 2023

CITATION

Deng H, Deng D, Qi T, Liu Z, Wu L and Yuan J (2023), An IFN- γ -related signature predicts prognosis and immunotherapy response in bladder cancer: Results from realworld cohorts. *Front. Genet.* 13:1100317. doi: 10.3389/fgene.2022.1100317

COPYRIGHT

© 2023 Deng, Deng, Qi, Liu, Wu and Yuan. This is an open-access article distributed under the terms of the Creative Commons Attribution License

(CC BY). The use, distribution or reproduction in other forums is permitted, provided the original author(s) and the copyright owner(s) are credited and that the original publication in this journal is cited, in accordance with accepted academic practice. No use, distribution or reproduction is permitted which does not comply with these terms.

An IFN-γ-related signature predicts prognosis and immunotherapy response in bladder cancer: Results from real-world cohorts

Hao Deng^{1,2†}, Dingshan Deng^{1,2†}, Tiezheng Qi³, Zhi Liu^{1,2}, Longxiang Wu^{1,2}* and Junbin Yuan^{1,2}*

¹Department of Urology, Xiangya Hospital, Central South University, Changsha, China, ²National Clinical Research Center for Geriatric Disorders, Xiangya Hospital, Central South University, Changsha, China, ³Xiangya School of Medicine, Central South University, Changsha, China

Bladder cancer (BLCA) is featured with high incidence and mortality. Whether the IFN- γ signaling could be used as an immunotherapy determinant for BLCA has not been fully confirmed. In this study, the transcriptome data and clinical information of BLCA samples were collected from The Cancer Genome Atlas (TCGA). Besides, four immunotherapy cohorts including IMvigor210 cohort, Gide cohort, Van Allen cohort, and Lauss cohort were collected. The Xiangya real-world cohort was used for independent validation. An IFN- γ -related signature was developed and validated in BLCA for predicting prognosis, mutation, tumor microenvironment status, and immunotherapy response. This is the first study focusing on the comprehensive evaluation of predictive values on the IFN- γ -related signature in BLCA. The potential clinical application of the IFN- γ -related signature was expected to be further validated with more prospective clinical cohorts.

KEYWORDS

bladder cancer, immunotherapy, IFN-y, prognosis, real-world cohort study

Introduction

Bladder cancer (BLCA) is recognized as one of the most common and heterogeneous urinary carcinomas worldwide (Antoni et al., 2017; Sung et al., 2021). Clinical data ceaselessly confirmed a high incidence and mortality in BLCA patients (Martinez Rodriguez et al., 2017). Thus, for decades, urologists have explored the mysteries of bladder cancer in hope of getting optimal solutions for precision BLCA treatments. Despite surgeries, radiotherapy, neoadjuvant or adjuvant chemotherapy and targeted therapy, BLCA patients still suffer (Antoni et al., 2017; Lenis et al., 2020). Most suffering patients are not sensitive to the current and mainstream treatment methods according to poor clinical outcomes (Lenis et al., 2020; Patel et al., 2020). Therefore, inventing new medical tools and treatment modalities for BLCA patients are urgently needed.

Cancer immunotherapy is a relatively burgeoning section in the field of cancer treatment, providing opportunities for alleviating, even curing BLCA patients (Pettenati and Ingersoll, 2018; Lenis et al., 2020). The immune checkpoint blockade (ICB), considered as the main direction of immunotherapy development, has been observed effective survival benefits in solid cancers, including BLCA (Rosenberg et al., 2016). Further from the cellular and molecular level, the ICB response rate mainly depends on the BLCA tumor microenvironment (TME) (Petitprez et al., 2020; Cao et al., 2021; Marin-Acevedo et al., 2021). TME is chiefly composed of cancer cells and immune cells, with other cell subsets and extracellular matrix as well (Hinshaw and Shevde, 2019). However, a malignant BLCA TME generally leads ICB moving towards failure. Potential mechanisms influencing ICB could be including the exhaustion and senescence of CD8 T cells in TME (Lian et al., 2020; Zhang et al., 2021); the huge secretion immunosuppressive factors (Metelli et al., 2018; Lecker et al., 2021). Besides, high tumor mutation burden (TMB), which could represent a high level of neoantigen, is a prominent characteristic of BLCA indicating a potential immunogenic microenvironment (Kandoth et al., 2013). Even though the response of patients with high TMB to ICB is considered heterogeneous, in BLCA, TMB is acknowleged to be a potential indicator reflecting ICB response efficiency (Chan et al., 2019; Liu et al., 2019). However, only a minor patients received ICB therapy achieved ideal outcomes. Therefore, exploring novel biomarkers for distinguishing specific groups of BLCA patients with an inflamed TME is necessary for increasing the ICB response rates.

IFN-γ signaling has been well-recognized as a critical mediator of tumor cell immunogenicity, which could help promote recognize and eliminate tumor cells (Dighe et al., 1994). IFN-γ expression has been proven to potentially predict clinical outcomes for multiple cancer types (Fridman et al., 2012), and is associated with mortality and disease risk of BLCA as well (Gillezeau et al., 2022). Besides, IFN-γ-induced was reported to up-regulate PD-ECGF/TP and enhance the cytotoxicity of 5-fluorouracil and 5'-deoxy-5-fluorouridine in BLCA (Li et al., 2002). Notably, IFN-γ-induced cytotoxicity has been revealed as a biomarker of resistance in BLCA (Green et al., 2021). However, whether the IFN-γ signaling could be used as an immunotherapy determinant for BLCA has not been fully confirmed.

In this study, an IFN- γ -related signature was developed in BLCA for predicting prognosis, mutation, tumor microenvironment, and immunotherapy. To date, we come first to comprehensively evaluate the predictive values of IFN- γ -related signature in BLCA.

Materials and methods

Data collection and procession

The transcriptome data and clinical information of BLCA samples were collected from The Cancer Genome Atlas (TCGA). The FPKM values of the raw matrix were transformed into TPM values for follow-up studies. The TCGA BLCA dataset included 403 BLCA samples and 19 normal samples. Four immunotherapy cohorts were collected, including IMvigor210 (248 samples), Gide (32 samples), Van Allen (42 samples) and Lauss (25 samples). Besides, the copy number variation (CNV) data of BLCA samples in the TCGA BLCA dataset, processed with GISTIC 2.0, were downloaded from the UCSC Xena data portal (http://xena.ucsc.edu/). Drug information, including 184 common anticancer drugs and the corresponding target genes were collected from the DrugBank database (www. drugbank.ca).

The Xiangya real-world cohort

According to our previous studies, the Xiangya real-world cohort was based on BLCA samples after surgical resections in the Xiangya Hospital, Central South University. The Xinagya real-world cohort, already uploaded with the number as GSE188715, included 57 BLCA samples sequenced by the BGISEQ-500 platform (BGI-Shenzhen, China) (Liu et al., 2021a; Hu et al., 2021). The TPM values of the raw matrix were used for the follow-up analysis.

Development of the IFN- γ -related signature in TCGA BLCA cohort

IFN-y related genes have never been systematically summarized as a list. Originally, we collected IFN-y-related genes from previous studies as comprehensively as possible (Gao et al., 2016; Hu et al., 2020). The least absolute shrinkage and selector operation (LASSO) regression analysis was first performed on these genes for dimension reduction. Each gene can be regarded as a factor, and factors which contribute relatively less to final outcomes of the analysis were assigned the value zero. Remaining factors, genes with nonzero coefficients, were ultimately selected for multivariable Cox regression analysis and used as variables to construct the IFN-y-related signature. The formula is as follows: Risk score = $\sum IFNi^*RNAi$, where IFNi is the coefficient of the genes and RNAi is the expression value of the genes in multivariable Cox regression analysis. To evaluate the of the IFN-γ-related signature, performance we simultaneously perform the time-dependent receiver operating characteristic (ROC) and calibration curves using



The expression pattern of the IFN- γ -related genes in the TCGA BLCA dataset. (A) Heatmap depicting the expression differences of the IFN- γ related genes between BLCA samples and normal samples. (B) Volcano plot for the expression differences of the IFN- γ -related genes between BLCA samples and normal samples. (C) 3D PCA plot depicting the expression differences of the IFN- γ -related genes between BLCA samples and normal samples. (D) Correlations between the IFN- γ -related genes. The size of the circle, calculated by the log-rank test and ranging from .1 to .0001, represents the prognosis of each gene. Green dots represent favorable factors, while purple dots represent risk factors. The color of the lines represents the correlation between the IFN- γ -related genes. Blue represents a negative correlation, and red represents a positive correlation. (E) The CNV frequency of the IFN- γ -related genes. The column represents the count, and the color represents gains or losses. Red represents gains, and blue represents losses. (F) The mutation frequency of the IFN- γ -related genes. The upper bar plot represents TMB. The number on the right represents the mutation frequency in each IFN- γ -related gene. The right bar plot represents the proportion of each variant type.

the R package "timeROC" in the TCGA BLCA cohort, Xiangya real-world cohort, and IMvigor210 cohort.

Development of a nomogram

The univariate and multivariate Cox regression analyses were adopted to filtrate independent prognostic factors from

clinicopathologic characteristics and the IFN- γ -related signature. After analyzing results, prognostic factors in univariate Cox analysis were screened and integrated to construct a nomogram. The performance of the nomogram was evaluated by the time-dependent ROC and calibration curves using the R package "timeROC" in the TCGA BLCA dataset and the Xiangya real-world cohort.



regression analysis. (B) The coefficients of the IFN- γ -related genes in LASSO regression analysis. (C) The coefficients of the IFN- γ -related genes in LASSO regression analysis. (D) The distribution of the IFN- γ -related signature in BLCA samples. (E) The survival curves of the two IFN- γ -related signature score groups. (F) 1-year, 3-year, and 5-year ROC of the IFN- γ -related signature.

The mutation landscape of the IFN- $\gamma\text{-}$ related signature

The mutation landscape of the IFN- γ -related signature was visualized based on CNV data using the R package "maftools." Somatic mutation data of BLCA samples in the TCGA BLCA dataset was used to calculate the tumor mutation burden (TMB).

The immunological characteristics of the $\ensuremath{\mathsf{IFN}}\xspace-\gamma\xspace$ related signature

In our previous studies, relevant immunological characteristics and algorithms in the TME were described in detail (Liu et al., 2021a; Hu et al., 2021). The cancer immunity cycle is composed of seven key steps: the release and presentation of cancer cell antigens (Steps 1 and 2), the priming and activation

of the immune system (Step 3), then the trafficking and infiltration of immune cells into tumors (Steps 4 and 5), finally recognizing and killing cancer cells by T cells (Steps 6 and 7) (26). The cancer immunity cycle and immune infiltrating cells involved were quantified using the single-sample gene-set enrichment analysis (ssGSEA) in BLCA samples.

Drug sensitivity and immunotherapy response prediction within the IFN- γ -related signature

We calculated the sensitivity of anticancer drugs using data downloaded from the DrugBank. The predictive value of the IFN- γ -related signature for the immunotherapy was validated in three immunotherapy cohorts, namely the Gide, Van Allen, and Lauss.

Statistical analysis

Correlation coefficients were determined by the Spearman and distance correlation analyses. Normally distributed continuous variables between the two groups were compared using the *t*-test, while the non-normally ones between the two groups were compared using the Mann-Whitney U test instead. Chi-square or Fisher exact tests were used for comparison between categorical variables. The "survcutpoint" function from the R package "survminer" for the maximum rank statistic was used to determine the optimal cutoff value of the IFN- γ -related signature. The survival curves were generated using the Kaplan-Meier method, while the statistical significance was determined using the log-rank test. The threshold of significance was set at p < .05, and all statistical tests were two-sided. R project (version 3.6.3, http://www.r-project.org) was used for all analyses.

Results

The expression pattern of the IFN- γ -related genes in the TCGA BLCA dataset

Differential analysis performed on the IFN- γ -related genes between BLCA samples and normal samples was shown in Figures 1A, B. Both sample sets (the BLCA set and normal one) could be clearly separated by IFN- γ -related genes (Figure 1C). Correlations between every two IFN- γ -related genes were shown in Figure 1D, in which 22 IFN- γ -related genes were favorable factors, while the remaining 11 were risk factors. The CNV frequency of IFN- γ -related genes was shown in Figure 1E. Roughly a half of the genes showed CNV loss, while the other half reversed. The mutation frequency of the IFN- γ -related genes was shown in Figure 1F, in which PARP14, OAS2, OAS3, C1S, and TNFAIP3 were the top five mutated IFN- γ -related genes.

Development of the IFN- γ -related signature in the TCGA BLCA dataset

We successfully reduced the dimension of IFN- γ -related genes using LASSO regression analysis (Figure 2A). The coefficients of the IFN- γ -related genes in LASSO regression analysis were shown in Figure 2B. TNFAIP2, CXCL10, and TAP1 were finally included for the multivariable Cox regression analysis. Figure 2C displayed the coefficients of these three genes. The distribution of the IFN- γ -related signature in BLCA samples was shown in Figure 2D. BLCA patients with high IFN- γ -related signature scores were associated with decreased survival time (Figure 2E). The accuracy of the signature in predicting 1-year, 3-year, and 5-year OS was .609, .614, and .639 respectively (Figure 2F).

Development of a nomogram

Univariate Cox regression analysis was performed on clinical characters including the IFN-y-related signature in the TCGA BLCA cohort. Results revealed that the age, tumor stage, TN grading system, and the IFN-y-related signature were independent prognostic factors (Figure 3A). Multivariate Cox regression analysis was ultimately performed likewise in the TCGA BLCA dataset, in which age and the IFN-y-related signature were independent prognostic factors (Figure 3B). Nomogram was constructed based on clinical factors including the IFN-y-related signature in the TCGA BLCA dataset (Figure 3C).1-year, 3-year, and 5-year ROC of the nomogram in the TCGA BLCA dataset had respective values of .72, .71, and .74 (Figure 4A). 1-year, 3year, and 5-year calibration curves of the nomogram in the TCGA BLCA dataset were shown in Figure 4B. 1year, 3-year, and 5-year ROC of the nomogram in the Xiangya real-world cohort had values of .82, .87, and .86 (Figure 4C). 1-year, 3-year, and 5-year calibration curves of the nomogram in the Xiangya real-world cohort were shown in Figure 4D.

The mutation landscape of the IFN- γ -related signature in the TCGA BLCA dataset

TP53, TTN, KMT2D, MUC16, and KDM6A were the top five mutated genes in the high IFN-γ-related signature score



group (Figure 5A). TTN, TP53, MUC16, KMT16, and ARID1A were the top five mutated genes in the low IFN-γ-related signature score group (Figure 5B). The high IFN-γ-related signature score group was significantly associated with a lower TMB level (Figure 5C). However, there was no significant correlation between the MANTIS score and the IFN- γ -related signature (Figure 5D).



FIGURE 4

The prognostic value of the nomogram. (A) 1-year, 3-year, and 5-year ROC of the nomogram in the TCGA BLCA dataset. (B) 1-year, 3-year, and 5-year calibration curves of the nomogram in the TCGA BLCA dataset. (C) 1-year, 3-year, and 5-year ROC of the nomogram in the Xiangya real-world cohort. (D) 1-year, 3-year, and 5-year calibration curves of the nomogram in the Xiangya real-world cohort.

The immunological characteristics of the IFN-y-related signature in the TCGA BLCA dataset

As was shown in Figure 6A, low IFN-y-related signature scores significantly indicated some cancer immunity cycles including T cell recruiting, Th 1 cell recruiting, and macrophage recruiting. Besides, the low IFN-y-related signature scores were generally associated with immune infiltrating cells including activated CD4 cells, activated CD8 cells, and natural killer T cells (Figure 6B). Correlations between the IFN-y-related signature and each stroma-activated pathway were shown in Figure 6C. Immunotherapy-predicted pathways were relatively more active in the low IFN-y-related signature score group (Figure 6D). We divided TCGA samples into different binary groups according to the sex and stage. Results of validating our IFN-y signature in female and male groups,

high stage and low stage groups proved the conclusion as expected (Supplementary Figure S1-S4).

The immunological characteristics of the IFN- γ -related signature in the Xiangya real-world cohort

BLCA patients with a high IFN-y-related signature score were associated with shorter survival time (Figure 7A). 1-year, 3year, and 5-year ROC of the IFN-y-related signature had values of .84, .66, and .66, respectively (Figure 7B). The signature also showed negative correlation with multiple immune checkpoint molecules, including CD274, LAG3, CTLA4, PDCD1, and HAVCR2 (Figure 7C). As expected, the signature was negatively associated with cancer immunity cycles, including T cell recruiting, Th 1 cell recruiting, and macrophage recruiting (Figure 7D). The IFN-y-related signature was



negatively associated with immunotherapy-predicted pathways, including APM signal, microRNAs in cancer, mismatch repair, cell cycle, and p53 signaling pathway (Figure 7D). In addition, the low signature score group was generally significantly associated with immune infiltrating cells, including activated CD4 cells, activated CD8 cells, and natural killer T cells (Figure 7E).

The immunological characteristics of the IFN- γ -related signature in the IMvigor210 cohort

BLCA patients with high IFN- γ -related signature scores were associated with decreased survival time (Figure 8A). 1year, 3-year, and 5-year ROC of the IFN- γ -related signature had values of .74, .59, and .6 (Figure 8B). The IFN-γ-related signature was negatively associated with multiple immune checkpoint molecules, including CD274, LAG3, CTLA4, PDCD1, and HAVCR2 (Figure 8C). The IFN-γ-related signature was negatively associated with cancer immunity cycles, including T cell recruiting, Th 1 cell recruiting, and macrophage recruiting (Figure 8D). The IFN-γ-related signature was negatively associated with immunotherapypredicted pathways, including APM signal, microRNAs in cancer, mismatch repair, cell cycle, and p53 signaling pathway (Figure 8D). In addition, the low IFN-γ-related signature scores were generally significantly associated with immune infiltrating cells, including activated CD4 cells, activated CD8 cells, and natural killer T cells (Figure 8E).



FIGURE 6

The immunological characteristics of the IFN- γ -related signature in the TCGA BLCA dataset. (A) Cancer immunity cycles between the two IFN- γ -related signature score groups. (B) Immune infiltrating cells between the two IFN- γ -related signature score groups. (C) Correlations between the IFN- γ -related signature and stroma-activated pathways. (D) Immunotherapy-predicted pathways between the two IFN- γ -related signature score groups. The left bar represents log10 *p*-values, the red bar represents activated pathways, and the blue bar represents inhibited pathways. (ns, Not Significant; **p* < .05; ***p* < .01; ****p* < .001; *****p* < .001).



The immunological characteristics of the IFN- γ -related signature in the Xiangya real-world cohort. (A) The survival curves of the two IFN- γ -related signature score groups. (B) 1-year, 3-year, and 5-year ROC of the IFN- γ -related signature. (C) Correlations between the IFN- γ -related signature and immune checkpoint molecules. (D) Correlations between the IFN- γ -related signature and cancer immunity cycles and immunotherapy-predicted pathways. (E) Correlations between the IFN- γ -related signature and immune infiltrating cells.



FIGURE 8

The immunological characteristics of the IFN- γ -related signature in the IMvigor210 cohort. (A) The survival curves of the two IFN- γ -related signature score groups. (B) 1-year, 3-year, and 5-year ROC of the IFN- γ -related signature. (C) Correlations between the IFN- γ -related signature and immune checkpoint molecules. (D) Correlations between the IFN- γ -related signature and cancer immunity cycles and immunotherapy-predicted pathways. (E) Correlations between the IFN- γ -related signature and immune infiltrating cells.



Drug prediction and immunotherapy response prediction of the IFN-γ-related signature. (A) The sensitivity of the anticancer drugs in the two IFN-γ-related signature score groups. The predictive value of the IFN-γ-related signature for the immunotherapy was validated in three immunotherapy cohorts, including. (B–D) Response of melanoma patients with high and low IFN-γ-related signature scores to ICB in Gide, Van Allen and Lauss respectively.

Drug prediction and immunotherapy response prediction of the IFN- γ -related signature

We predicted the sensitivity of BLCA patients to common drugs. Atezolizumab_CD274, Cetuximab_FCGR2A, Cetuximab_FCGR3A, Cetuximab_C1QC, Cetuximab_C1QB, Cetuximab_FCGR1A, Cetuximab_G1QA, Trastuzumab_ ERBB2, and Bevacizumab_VEGFA had significantly lower drug sensitivity in the high IFN- γ -related signature score group (Figure 9A). Likewise, melanoma patients with high IFN- γ -related signature scores were less likely to respond to ICB (Figures 9B–D). Notably, melanoma patients with high IFN- γ -related signature scores were associated with decreased survival time (Figure 9C).

Discussion

Increasing evidence constantly confirmed that IFN-y plays a critical role in the tumorigenicity and immunogenicity of various cancers (Jorgovanovic et al., 2020; Stifter et al., 2020). Specially, IFNy is a cytokine that physiologically promotes innate and adaptive immune responses, while preventing the development of primary and transplanted tumors (Burke and Young, 2019). However, the potential roles of IFN-y in the prognosis and especially in the TME of BLCA remain unclear. Recently, mining markers precisely predicting prognosis and survival in cancers based on large-scale bioinformatic analysis has received much more attention in the big data era than ever before. Several markers have been proven robust in predicting survival outcomes and immunotherapy responses (Zhang et al., 2022a; Zhang et al., 2022b). Thus, we aimed to explore the predictive value of IFN-y and its related genes in BLCA using comprehensive bioinformatics on internal datasets and external real-world validation cohort.

Our study originally proved that most IFN-γ-related genes were risk factors in BLCA.

After summarizing IFN-γ-related genes, a corresponding signature was successfully developed in TCGA BLCA cohort. The predictive signature consisted of three genes, namely TNFAIP2, CXCL10, and TAP1. From with, TNFAIP2, a primary response gene of TNFα, is highly expressed in immune cells and the urinary bladder cells (Jia et al., 2016; Niwa et al., 2019). TNFAIP2 has been reported to promote proliferation, angiogenesis, migration, and invasion of cancer (Jia et al., 2018). CXCL10, CXCL9, CXCL11/CXCR3 is an important axis for immune activation, which is necessary for developing novel cancer therapy (Tokunaga et al., 2018). In addition, a significantly negative correlation between TAP1 and survival in breast, lung, liver, and ovarian cancer is revealed (Tabassum et al., 2021).

The IFN-y-related signature could predict the survival outcomes of BLCA patients in the TCGA BLCA dataset, Xiangya real-world cohort and IMvigor210 cohort. Cox regression analysis determined that the signature was an independent prognostic factor as age, tumor stage, and TN grading system in the predictive nomogram. The constructed nomogram performs robustly in predicting the survival outcomes of BLCA patients. As generally agreed, the TNM staging system is the most widely accepted and most commonly used system for BLCA (Adsay et al., 2012). The IFN-y-related signature showed sensationally superior performance compared to the TNM staging system regarding predicting the prognosis of BLCA patients. TP53 is an important tumor suppressor gene that is frequently mutated in cancer (Donehower et al., 2019). While MUC16 and TTN genes mutation were previously reported to correlate with prognosis, tumor mutation burden, and immunotherapy efficacy in cancers (Yang et al., 2020). KMT2D deficiency was found to impair super-enhancers to confer a glycolytic vulnerability in lung cancer (Alam et al., 2020). KDM6A-ARHGDIB axis could block metastasis of BLCA by inhibiting Rac1 (Liu et al., 2021b). In accordance with these

findings, TP53, TTN, KMT2D, MUC16, and KDM6A were the top five mutated genes in the high IFN- γ -related signature score group.

The core part of immunotherapy is to help the immune system recognize and destroy tumor cells through enhancing the reaction of immune cells to present tumor antigens (Frankel et al., 2017). TME, composed of cancer cells, non-cancerous cells (mainly immune infiltrating cells), and secreted cytokines, has emerged as a promising mediator for immunotherapy (Bejarano et al., 2021). An immune hot TME, which is infiltrated by more immune cells, is more likely to present a better ICB response (Wang et al., 2022). The IFN-y-related signature was negatively associated with cancer immunity cycles, immunotherapy-predicted pathways, and immune infiltrating cells in the TCGA BLCA dataset, Xiangya real-world cohort, and IMvigor210 cohort. As the most important determinant for ICB, immune checkpoint molecules have been widely studied in the past few decades. CD274, LAG3, CTLA4, PDCD1, and HAVCR2 have been the most promising immune checkpoint molecules with satisfying results in clinical trials (Wang et al., 2022). As expected, the IFN-y-related signature was negatively associated with multiple immune checkpoint molecules, including CD274, LAG3, CTLA4, PDCD1, and HAVCR2. Besides, the high IFN-yrelated signature score group was significantly associated with a lower TMB level. These results indicate potential lower response rates in BLCA patients with high IFN-y-related signature scores. The direct immunotherapy response prediction of the IFN-y-related signature in Gide, Van Allen, and Lauss cohorts proved this finding. Furthermore, IFN-y-related signature could predict drug sensitivity of Atezolizumab_CD274, Cetuximab_FCGR2A, Cetuximab_FCGR3A, Cetuximab_C1QC, Cetuximab_C1QB, Cetuximab_FCGR1A, Cetuximab_G1QA, Trastuzumab_ERBB2, and Bevacizumab_VEGFA.

To sum up, an IFN- γ -related signature was developed in BLCA for predicting prognosis, mutation, tumor microenvironment, and immunotherapy. The potential clinical application of the IFN- γ -related signature is expected to be further validated by more clinical cohorts.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/ Supplementary Material.

Ethics statement

The studies involving human participants were reviewed and approved by the Ethics Committee of the Xiangya Hospital of Central South University. The patients/participants provided their written informed consent to participate in this study. The animal study was reviewed and approved by the Ethics Committee of the Xiangya Hospital of Central South University. Written informed consent was obtained from the individual(s) for the publication of any potentially identifiable images or data included in this article.

Author contributions

Conception and design: LW, JY, HD, and DD. Provision of study materials or patients: TQ and ZL. Collection and assembly of data: TQ and ZL. Data analysis and interpretation: HD, DD, TQ, and ZL. Manuscript writing: HD, DD, TQ, ZL, LW, and JY. Final approval of manuscript: All authors.

Funding

This work was supported by the grants from the Natural Science Foundation of Hunan Province (2021JJ31122).

Acknowledgments

We sincerely thank all participants in the study.

Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

References

Adsay, N. V., Bagci, P., Tajiri, T., Oliva, I., Ohike, N., Balci, S., et al. (2012). Pathologic staging of pancreatic, ampullary, biliary, and gallbladder cancers: Pitfalls and practical limitations of the current AJCC/UICC TNM staging system and opportunities for improvement. *Semin. Diagn Pathol.* 29, 127–141. doi:10.1053/j.semdp.2012.08.010

Alam, H., Tang, M., Maitituoheti, M., Dhar, S. S., Kumar, M., Han, C. Y., et al. (2020). KMT2D deficiency impairs super-enhancers to confer a glycolytic vulnerability in lung cancer. *Cancer Cell* 37, 599–617. doi:10.1016/j.ccell.2020.03.005

Antoni, S., Ferlay, J., Soerjomataram, I., Znaor, A., Jemal, A., and Bray, F. (2017). Bladder cancer incidence and mortality: A global overview and recent trends. *Eur. Urol.* 71, 96–108. doi:10.1016/j.eururo.2016.06.010

Bejarano, L., Jordao, M. J. C., and Joyce, J. A. (2021). Therapeutic targeting of the tumor microenvironment. *Cancer Discov.* 11, 933–959. doi:10.1158/2159-8290.cd-20-1808

Burke, J. D., and Young, H. A. (2019). IFN-F: A cytokine at the right time, is in the right place. *Semin. Immunol.* 43, 101280. doi:10.1016/j.smim.2019.05.002

Cao, R., Yuan, L., Ma, B., Wang, G., and Tian, Y. (2021). Tumour microenvironment (TME) characterization identified prognosis and immunotherapy response in muscle-invasive bladder cancer (MIBC). *Cancer Immunol. Immunother.* 70, 1–18. doi:10.1007/s00262-020-02649-x

Chan, T. A., Yarchoan, M., Jaffee, E., Swanton, C., Quezada, S. A., Stenzinger, A., et al. (2019). Development of tumor mutation burden as an immunotherapy biomarker: Utility for the oncology clinic. *Ann. Oncol.* 30, 44–56. doi:10.1093/annonc/mdy495

Dighe, A. S., Richards, E., Old, L. J., and Schreiber, R. D. (1994). Enhanced *in vivo* growth and resistance to rejection of tumor cells expressing dominant negative IFN gamma receptors. *Immunity* 1, 447–456. doi:10.1016/1074-7613(94)90087-6

Publisher's note

All claims expressed in this article are solely those of the authors and do not necessarily represent those of their affiliated organizations, or those of the publisher, the editors and the reviewers. Any product that may be evaluated in this article, or claim that may be made by its manufacturer, is not guaranteed or endorsed by the publisher.

Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fgene. 2022.1100317/full#supplementary-material

SUPPLEMENTARY FIGURE S1

Exploring correlations between the IFN- γ score and TME characteristics in female group. (A) Differences in cancer immune cycle activity between high-risk and low-risk groups. (B) Correlation between the score and the infiltration of different immune cells. (C) Relationship between the TIS and the IFN- γ score. (D) Correlation between the IFN- γ score and various common ICB response pathways. (E) Correlation between the IFN- γ score and immune checkpoints.

SUPPLEMENTARY FIGURE S2

Exploring correlations between the $\mathsf{IFN}\text{-}\gamma$ score and TME characteristics in male group.

SUPPLEMENTARY FIGURE S3

Exploring correlations between the IFN- γ score and TME characteristics in high stage group.

SUPPLEMENTARY FIGURE S4

Exploring correlations between the IFN- γ score and TME characteristics in low stage group.

Donehower, L. A., Soussi, T., Korkut, A., Liu, Y., Schultz, A., Cardenas, M., et al. (2019). Integrated analysis of TP53 gene and pathway alterations in the cancer Genome Atlas. *Cell Rep.* 28, 3010–1384 e5. doi:10.1016/j.celrep.2019. 08.061

Frankel, T., Lanfranca, M. P., and Zou, W. (2017). The role of tumor microenvironment in cancer immunotherapy. *Adv. Exp. Med. Biol.* 1036, 51–64. doi:10.1007/978-3-319-67577-0_4

Fridman, W. H., Pages, F., Sautes-Fridman, C., and Galon, J. (2012). The immune contexture in human tumours: Impact on clinical outcome. *Nat. Rev. Cancer* 12, 298–306. doi:10.1038/nrc3245

Gao, J., Shi, L. Z., Zhao, H., Chen, J., Xiong, L., He, Q., et al. (2016). Loss of IFNgamma pathway genes in tumor cells as a mechanism of resistance to anti-CTLA-4 therapy. *Cell* 167, 397–404. doi:10.1016/j.cell.2016.08.069

Gillezeau, C., Movva, N., van Gerwen, M., Rabon-Stith, K., Shire, N., Brohawn, P. Z., et al. (2022). Interferon gamma expression and mortality in unselected cohorts of urothelial bladder cancer patients. *PLoS One* 17, e0271339. doi:10.1371/journal.pone. 0271339

Green, J. L., Osterhout, R. E., Klova, A. L., Merkwirth, C., McDonnell, S. R. P., Zavareh, R. B., et al. (2021). Molecular characterization of type I IFN-induced cytotoxicity in bladder cancer cells reveals biomarkers of resistance. *Mol. Ther. Oncolytics* 23, 547–559. doi:10.1016/j.omto.2021.11.006

Hinshaw, D. C., and Shevde, L. A. (2019). The tumor microenvironment innately modulates cancer progression. *Cancer Res.* 79, 4557–4566. doi:10.1158/0008-5472. CAN-18-3962

Hu, B., Wei, Q., Li, X., Ju, M., Wang, L., Zhou, C., et al. (2020). Development of an IFNγ response-related signature for predicting the survival of cutaneous melanoma. *Cancer Med.* 9, 8186–8201. doi:10.1002/cam4.3438

Hu, J., Yu, A., Othmane, B., Qiu, D., Li, H., Li, C., et al. (2021). Siglec15 shapes a non-inflamed tumor microenvironment and predicts the molecular subtype in bladder cancer. *Theranostics* 11, 3089–3108. doi:10.7150/thno.53649

Jia, L., Shi, Y., Wen, Y., Li, W., Feng, J., and Chen, C. (2018). The roles of TNFAIP2 in cancers and infectious diseases. *J. Cell Mol. Med.* 22, 5188–5195. doi:10.1111/jcmm.13822

Jia, L., Zhou, Z., Liang, H., Wu, J., Shi, P., Li, F., et al. (2016). KLF5 promotes breast cancer proliferation, migration and invasion in part by upregulating the transcription of TNFAIP2. *Oncogene* 35, 2040–2051. doi:10.1038/onc.2015.263

Jorgovanovic, D., Song, M., Wang, L., and Zhang, Y. (2020). Roles of IFN- γ in tumor progression and regression: A review. *Biomark. Res.* 8, 49. doi:10.1186/s40364-020-00228-x

Kandoth, C., McLellan, M. D., Vandin, F., Ye, K., Niu, B., Lu, C., et al. (2013). Mutational landscape and significance across 12 major cancer types. *Nature* 502, 333–339. doi:10.1038/nature12634

Lecker, L. S. M., Berlato, C., Maniati, E., Delaine-Smith, R., Pearce, O. M. T., Heath, O., et al. (2021). TGFBI production by macrophages contributes to an immunosuppressive microenvironment in ovarian cancer. *Cancer Res.* 81, 5706–5719. doi:10.1158/0008-5472.can-21-0536

Lenis, A. T., Lec, P. M., Chamie, K., and Mshs, M. D. (2020). Bladder cancer: A review. Jama 324, 1980–1991. doi:10.1001/jama.2020.17598

Li, G., Kawakami, S., Kageyama, Y., Yan, C., Saito, K., and Kihara, K. (2002). IFN gamma-induced up-regulation of PD-ECGF/TP enhances the cytotoxicity of 5-fluorouracil and 5'-deoxy-5-fluorouridine in bladder cancer cells. *Anticancer Res.* 22, 2607–2612.

Lian, J., Yue, Y., Yu, W., and Zhang, Y. (2020). Immunosenescence: A key player in cancer development. J. Hematol. Oncol. 13, 151. doi:10.1186/s13045-020-00986-z

Liu, L., Bai, X., Wang, J., Tang, X. R., Wu, D. H., Du, S. S., et al. (2019). Combination of TMB and CNA stratifies prognostic and predictive responses to immunotherapy across metastatic cancer. *Clin. Cancer Res.* 25, 7413–7423. doi:10. 1158/1078-0432.CCR-19-0558

Liu, L., Cui, J., Zhao, Y., Liu, X., Chen, L., Xia, Y., et al. (2021). KDM6A-ARHGDIB axis blocks metastasis of bladder cancer by inhibiting Rac1. *Mol. Cancer* 20, 77. doi:10.1186/s12943-021-01369-9

Liu, Z., Qi, T., Li, X., Yao, Y., Othmane, B., Chen, J., et al. (2021). A novel TGF- β risk score predicts the clinical outcomes and tumour microenvironment phenotypes in bladder cancer. *Front. Immunol.* 12, 791924. doi:10.3389/fimmu.2021.791924

Marin-Acevedo, J. A., Kimbrough, E. O., and Lou, Y. (2021). Next generation of immune checkpoint inhibitors and beyond. *J. Hematol. Oncol.* 14, 45. doi:10.1186/s13045-021-01056-8

Martinez Rodriguez, R. H., Buisan Rueda, O., and Ibarz, L. (2017). Bladder cancer: Present and future. *Med. Clin. Barc.* 149, 449–455. doi:10.1016/j.medcli.2017.06.009

Metelli, A., Salem, M., Wallace, C. H., Wu, B. X., Li, A., Li, X., et al. (2018). Immunoregulatory functions and the therapeutic implications of GARP-TGF- β in inflammation and cancer. *J. Hematol. Oncol.* 11, 24. doi:10.1186/s13045-018-0570-z

Niwa, N., Tanaka, N., Hongo, H., Miyazaki, Y., Takamatsu, K., Mizuno, R., et al. (2019). TNFAIP2 expression induces epithelial-to-mesenchymal transition and

confers platinum resistance in urothelial cancer cells. Lab. Invest 99, 1702–1713. doi:10.1038/s41374-019-0285-y

Patel, V. G., Oh, W. K., and Galsky, M. D. (2020). Treatment of muscle-invasive and advanced bladder cancer in 2020. *CA Cancer J. Clin.* 70, 404–423. doi:10.3322/caac.21631

Petitprez, F., Meylan, M., de Reyniès, A., Sautès-Fridman, C., and Fridman, W. H. (2020). The tumor microenvironment in the response to immune checkpoint blockade therapies. *Front. Immunol.* 11, 784. doi:10.3389/fimmu.2020.00784

Pettenati, C., and Ingersoll, M. A. (2018). Mechanisms of BCG immunotherapy and its outlook for bladder cancer. *Nat. Rev. Urol.* 15, 615–625. doi:10.1038/s41585-018-0055-4

Rosenberg, J. E., Hoffman-Censits, J., Powles, T., van der Heijden, M. S., Balar, A. V., Necchi, A., et al. (2016). Atezolizumab in patients with locally advanced and metastatic urothelial carcinoma who have progressed following treatment with platinum-based chemotherapy: A single-arm, multicentre, phase 2 trial. *Lancet* 387, 1909–1920. doi:10.1016/S0140-6736(16)00561-4

Stifter, K., Krieger, J., Ruths, L., Gout, J., Mulaw, M., Lechel, A., et al. (2020). IFN- γ treatment protocol for MHC-1(lo)/PD-L1(+) pancreatic tumor cells selectively restores their TAP-mediated presentation competence and CD8 T-cell priming potential. *J. Immunother. Cancer* 8, e000692. doi:10. 1136/jitc-2020-000692

Sung, H., Ferlay, J., Siegel, R. L., Laversanne, M., Soerjomataram, I., Jemal, A., et al. (2021). Global cancer statistics 2020: GLOBOCAN estimates of incidence and mortality worldwide for 36 cancers in 185 countries. *CA Cancer J. Clin.* 71, 209–249. doi:10.3322/caac.21660

Tabassum, A., Samdani, M. N., Dhali, T. C., Alam, R., Ahammad, F., Samad, A., et al. (2021). Transporter associated with antigen processing 1 (TAP1) expression and prognostic analysis in breast, lung, liver, and ovarian cancer. *J. Mol. Med. Berl.* 99, 1293–1309. doi:10.1007/s00109-021-02088-w

Tokunaga, R., Zhang, W., Naseem, M., Puccini, A., Berger, M. D., Soni, S., et al. (2018). CXCL9, CXCL10, CXCL11/CXCR3 axis for immune activation - a target for novel cancer therapy. *Cancer Treat. Rev.* 63, 40–47. doi:10.1016/j.ctrv.2017.11.007

Wang, Y., Zhang, H., Liu, C., Wang, Z., Wu, W., Zhang, N., et al. (2022). Immune checkpoint modulators in cancer immunotherapy: Recent advances and emerging concepts. *J. Hematol. Oncol.* 15, 111. doi:10.1186/s13045-022-01325-0

Yang, Y., Zhang, J., Chen, Y., Xu, R., Zhao, Q., and Guo, W. (2020). MUC4, MUC16, and TTN genes mutation correlated with prognosis, and predicted tumor mutation burden and immunotherapy efficacy in gastric cancer and pan-cancer. *Clin. Transl. Med.* 10, e155. doi:10.1002/ctm2.155

Zhang, H., Zhang, N., Wu, W., Zhou, R., Li, S., Wang, Z., et al. (2022). Machine learning-based tumor-infiltrating immune cell-associated lncRNAs for predicting prognosis and immunotherapy response in patients with glioblastoma. *Brief. Bioinform* 23, bbac386. doi:10.1093/bib/bbac386

Zhang, J., He, T., Xue, L., and Guo, H. (2021). Senescent T cells: A potential biomarker and target for cancer therapy. *EBioMedicine* 68, 103409. doi:10.1016/j. ebiom.2021.103409

Zhang, N., Zhang, H., Wu, W., Zhou, R., Li, S., Wang, Z., et al. (2022). Machine learning-based identification of tumor-infiltrating immune cellassociated lncRNAs for improving outcomes and immunotherapy responses in patients with low-grade glioma. *Theranostics* 12, 5931–5948. doi:10.7150/ thno.74281