Introduction

Pancreatic cancer is the seventh leading cause of cancer-related death in the world, with a 5-year survival rate of only 9% (1). One of the significant reasons for the high mortality is that pancreatic cancer is difficult to diagnose in an early stage, and 80–90% of patients have missed the best opportunity for surgery when they were diagnosed (2). Patients with pancreatic cancer frequently present with...
nonspecific symptoms such as abdominal pain and weight loss, which can delay diagnosis (3). Pancreatic cancer currently lacks specific methods for early diagnosis.

There is no reliable screening test currently available to screen the general population and detect pancreatic cancer early (4,5). The main tumor markers for clinical diagnosis of pancreatic cancer are carbohydrate antigen 19-9 (CA19-9) and carbohydrate antigen 242 (CA242). However, their sensitivity and specificity are not very satisfying (6-11). CA19-9 is the most commonly used serum-based marker for diagnosis of pancreatic cancer (12,13). However, this biomarker has some limitations. The level of CA19-9 can be normal in patients with localized disease, so the effect of early screening pancreatic carcinoma (PAAD) was not obvious. High CA19-9 levels can also take place in benign diseases, including chronic pancreatitis and benign jaundice (12-14).

Therefore, to explore specific tumor markers for early diagnosis of pancreatic cancer is the key to improve the overall therapeutic effect of patients with pancreatic cancer.

Another important reason for the poor 5-year survival rate is that pancreatic cancer is resistant to current treatment modalities, including chemotherapy, radiotherapy and targeted therapies (15). Immunotherapy has been shown to be effective in a variety of tumors, but pancreatic cancer remains completely resistant to it (16). It has also been reported that the tumor microenvironment critically influence the gene expression of tumor tissues and the clinical outcomes (17-22). An important feature of the tumor immune microenvironment in pancreatic cancer is that there are abundant non-cancer cell components in pancreatic cancer tissue, which constitute the tumor matrix. The tumor matrix accounts for more than 50% of the total tumor mass. Current studies have shown that the pancreatic tumor microenvironment has a high degree of immunosuppression and inhibits the body’s spontaneous and treatment-induced antitumor immunity (23).

Pumilio homologous protein 1 (PUM1) is a sequence-specific RNA binding protein (24,25) that is involved in various physiological processes, such as the cell cycle, DNA repair, and cell renewal (26-29). It has been proven that PUM1 plays an oncogene role in ovarian cancer, non-small-cell lung carcinoma (NSCLC), lymphocyte leukemia and other tumors (30-33). In our previous studies, we also found that the expression of PUM1 is correlated with the stage and prognosis of pancreatic cancer (34).

In this study, we used a variety of bioinformatic tools and analyzed data from patients with pancreatic cancer in The Cancer Genome Atlas (TCGA). We attempt to explore the relationship between expression of PUM1 and prognosis of patients with PAAD, the potential of PUM1 as a diagnostic marker for PAAD and its possible regulatory mechanisms. We present the following article in accordance with the MDAR reporting checklist (available at http://dx.doi.org/10.21037/tcr-20-3295).

Methods

Ethics statement

The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). The study was approved by Research Ethics Committee of Southwest Hospital, Army Medical University (KY2020138). All the datasets were retrieved from published literature, and written informed consent was confirmed in all studies.

Data preparation

Data on the expression profile of PUM1 mRNA and the corresponding clinical information of the pancreatic cancer samples and normal samples, as well as the tumor subgroup of cancer samples, were obtained from the TCGA (https://www.cancer.gov/about-nci/organization/ccg/research/structural-genomics/tcga) database and Genotype-Tissue Expression (GTEx) (https://gtexportal.org/) database. The TCGA has sequencing and pathological data for 30 different cancers (35). TCGA contained 9,736 carcinoma tissues, and GTEx contained RNA-seq data of 8,587 adjacent non-tumor tissues. The data from TCGA and GTEx are publicly available and open-access. This study follows the TCGA data access policy and published guidelines.

Gene Expression Profiling Interactive Analysis (GEPIA) and Human Protein Atlas (HPA) analysis

The HPA database (https://proteinforma.org/), Kaplan-Meier plotter survival analysis platform (http://kmplot.com/), and GEPIA database (http://gepia.cancer-pku.cn) were employed to examine PUM1 expression and survival in different cancers, including pancreatic cancer.

GEPIA is a newly developed interactive web server. It uses a standard processing pipeline to analyze the RNA-seq expression data of 9,736 tumors and 8,587 normal samples from the TCGA and GTEx projects. Customizable functions are provided by GEPIA, such as tumor/normal
differential expression analysis, profiling according to cancer types or pathological stages, patient survival analysis, similar gene detection, correlation analysis and dimensionality reduction analysis (36). In this study, using GEPIA database, 179 pancreatic cancer samples and 171 normal samples were collected to analyze the expression levels of PUM1 in pancreatic cancer.

**BBCancer analysis**

The difference in PUM1 mRNA levels of extracellular vesicles in the blood between patients with pancreatic cancer and healthy individuals was analyzed using the BBCancer (http://bbcancer.renlab.org/) database (37). The BBCancer database is derived from the Gene Expression Omnibus (GEO) (https://www.ncbi.nlm.nih.gov/geo/), miRBase V22 (http://www.mirbase.org/), piRNABank (http://pirnabank.ibab.ac.in/), and Minebase V2.0 (https://cm.jefferson.edu/MINTbase/) databases.

**cBioPortal and MethSurv analysis**

The analyses of PUM1 expression and methylation used cBioPortal (https://www.cbioportal.org/) (38). The genomic profiles involved mutations from GISTIC, Z-scores for mRNA expression (RNA-seq V2 RSEM) and Z-scores for protein expression (from reverse-phase protein array, RPPA). Co-expression and network were calculated on the basis of the online instructions of cBioPortal.

The relationship between the DNA methylation of PUM1 and survival was analyzed using the MethSurv database (https://biit.cs.ut.ee/methsurv/) (39). The MethSurv database is a web-based tool for multivariate survival analysis using DNA methylation data derived from the methylation group data of TCGA and developed using a Cox proportional hazard model, CpG, which allows survival analysis of CpGs located in or adjacent to a query gene and includes 7,358 methylation sites in 25 different cancer types.

**The Cancer Imaging Archive (TCIA) and xCell analysis**

Using the RNA-seq data of TCIA and xCell gene expression, the CIBERSORT algorithm was used to estimate the proportion of 22 infiltrating immune cells.

The TCIA (https://tcia.at/) includes tumors from 20 solid cancers and uses gene enrichment analysis and deconvolution to calculate the contents of individual immune cells (40). Gene expression of specific immune-related gene sets, composition of tumor-infiltrating immune cells (TIICs), new tumor antigens and cancer lineage antigens, HLA types and tumor heterogeneity can be found. The database is based on the features of infiltrating immune cells that respond to tumor genotypes that determine immunophenoscores (IPSs) and immune escape mechanisms.

xCell (https://xcell.ucsf.edu/) is a network tool that can perform cell type enrichment analysis on gene expression data of 64 immune cells and interstitial cells (41). xCell is a method based on genetic markers, with thousands of pure cell types. The results of xCell analysis were verified through extensive computer simulations and cellular immunophenotyping.

**Tumor Immune Estimation Resource (TIMER) database analysis**

TIMER is a comprehensive resource which can systematically analyze the immune infiltrates in various cancer types (https://cistrome.shinyapps.io/timer/) (42). A previously published statistical deconvolution method (43) is used by TIMER to infer the abundance of TIICs from gene expression profiles. The TIMER database covers 10,897 samples across 32 cancer types from the TCGA for estimating the abundance of immune infiltrates. We analyzed the expression of PUM1 in different types of cancer and the correlation between PUM1 expression and the abundance of immune infiltrates, including B cells, CD4+ T cells, CD8+ T cells, neutrophils, macrophages, and dendritic cells, via gene modules. The correlation module generated scatter plots of expression between a pair of user-defined genes in a given cancer type, and the Spearman's correlation and estimated statistical significance of the estimates. LAYN was used for the X-axis with gene symbols, and associated marker genes are represented as gene symbols on the Y-axis. The log2 RSEM is used to show gene expression levels.

**Diagnostic prediction**

According to the best cut-off value obtained from the receiver operating characteristic (ROC) curve, patients with pancreatic cancer were divided into high- and low-mRNA-expression groups. In addition, the area under the curve (AUC) of the ROC curve was computed to evaluate the predictive power of PUM1 in pancreatic cancer diagnosis. In the case of AUC >0.5, the closer the AUC is to 1, the
better the diagnosis.

**Statistical methods**

All results are presented as the mean ± standard deviation. Excel, GraphPad 7.0, and SPSS 25.0 (IBM, Chicago, IL, USA) were used to analyze the data. Comparisons between two groups were conducted using Student’s t-test. Comparisons among three or more groups were conducted using one-way analysis of variance (ANOVA). Independent-samples t-test was used to statistically analyze the measurement data. The survival data were analyzed using the Kaplan-Meier method. Pearson’s χ² method was used to determine the correlations between PUM1 expression and clinical parameters. P<0.05 was considered to be statistically significant.

**Results**

**PUM1 mRNA expression correlates with the prognosis of pancreatic cancer patients**

We obtained data from the GEPIA database and evaluated the transcription level of PUM1. The data revealed that the mRNA expression of PUM1 in diffuse large B-cell lymphoma, esophageal cancer, pancreatic cancer, gastric cancer and thymoma was significantly higher than that in the corresponding normal tissues. The expression of PUM1 mRNA in pancreatic cancer was 24.64 transcripts per million (TPM) and that in normal pancreas was 8.09 TPM (P<0.01). The fold difference was approximately three (Figure 1A). Figure 1B provides a scatter diagram of the difference in PUM1 mRNA expression between pancreatic tumors and normal tissues. Thus, PUM1 expression may serve as a potential diagnostic indicator in PAAD.

In our previous research, we verified that PUM1 expression in pancreatic cancer is correlated with the prognosis of patients (34). We then explored the prognostic significance of the PUM1 expression level in pancreatic cancer using the HPA database. Figure 1C shows the distribution of the PUM1 expression profiles in the high-expression (7%, n=37) and low-expression (38%, n=139) subgroups in pancreatic cancer. We calculated the best expression cut-off value (18.11%) according to the P score and median expression value to separate the patients into high-expression and low-expression subgroups (Figure 1D). Figure 1E shows that the survival probability of the high-expression group was significantly lower than that of the low-expression group (hazard ratio (HR) [95% confidence interval (CI)] =1.894 (1.131–3.170), log-rank test P value =0.0032). High PUM1 expression was associated with unfavorable prognosis in pancreatic cancer.

**The PUM1 mRNA expression level correlates with the incidence of pancreatic cancer**

The results of ROC curves showed that PUM1 mRNA expression levels correlated with pancreatic cancer incidence. Detailed ROC results are provided in Figure 2A (AUC value =0.9734, P<0.0001). Our results suggest that PUM1 mRNA levels in tissues can be exploited as useful biomarkers to diagnose pancreatic cancer.

We further analyzed the PUM1 mRNA of extracellular vesicles in the blood. The expression in the patients with pancreatic cancer (n=14) was more than 5 TPM, while that in normal people (n=6) was almost undetectable [log fold change (FC) =6.3264, P<0.0001, Figure 2B]. ROC curve analysis showed that the PUM1 mRNA level of extracellular vesicles in the blood is an excellent diagnostic marker for pancreatic cancer (AUC =1, P=0.0005, Figure 2C).

**Association of the DNA methylation level of PUM1 with the prognosis of pancreatic cancer patients**

What affects the expression level of PUM1 in pancreatic cancer? Here, we analyzed the association between gene expression and methyltransferase (DNMT) expression in diverse tumors using TCGA expression profile data. As shown in Figure 3A, PUM1 expression in PAAD was positively correlated with the four methyltransferases (DNMT1: red, r=0.52, P<0.0001; DNMT2: blue, r=0.6, P<0.0001; DNMT3A: green, r=0.46, P<0.0001; DNMT3B: purple, r=0.18, P=0.016). We further explored the association between the mRNA expression and methylation status of PUM1 to elucidate potential mechanisms of abnormal upregulation in pancreatic cancer tissues. Analysis of the data from TCGA in the cBioPortal database showed that PUM1 mRNA expression in pancreatic cancer was negatively correlated with methylation of PUM1 (r=–0.29, P=0.0001, Figure 3B). We then investigated the association between PUM1 methylation and patient prognosis using the MethSurv database. The analysis showed that the frequencies of detected sites with low, moderate and high methylation levels were 52.17% (12/23), 8.70% (2/23) and 47.83% (9/23), respectively (Figure 3C). We performed Kaplan-Meier plotter analysis based on the
Figure 1  PUM1 mRNA expression is correlated with the prognosis in patients with PAAD. (A) PUM1 mRNA expression profile in normal tissues and corresponding tumors. (B) The GEPIA was used to analyze the expression levels of PUM1 mRNA in PAAD tissues (tumor) and normal tissues (normal). *, P<0.05. (C) The distribution of the PUM1 expression profiles in the high-expression (7%, n=37) and low-expression (38%, n=139) subgroups in pancreatic cancer in the HPA database. (D) The best expression cut-off value is 18.11% according to the P score and dead median separation between high-expression and low-expression subgroups. (E) Survival curve based on PUM1 expression level from the Kaplan-Meier plotter survival analysis platforms [HR (95% CI) =1.894 (1.131–3.170), log-rank test P value =0.0032]. PUM1, pumilio homologous protein 1; PAAD, pancreatic adenocarcinoma; GEPIA, Gene Expression Profiling Interactive Analysis; HPA, Human Protein Atlas; HR, hazard ratio; CI, confidence interval.
methylation level profiles for the 23 specific sites and the prognosis information. Of a total of 23 methylation sites, 5 (cg04976330, cg23281075, cg04078732, cg0891023, cg08849613) were associated with survival outcomes, as shown in Figure 3D. The other 18 methylation sites are shown in the Figure S1. Hypomethylation was common among with high-risk patients, while hypermethylation was common among with low-risk patients.

**Association of PUM1 expression and the immune microenvironment in pancreatic cancer**

How does PUM1 play a role in pancreatic cancer? Analysis of data from the TCGA pancreatic cancer dataset with the xCell web tool showed that increased expression of PUM1 reduced the proportions of common lymphoid progenitors (CLP, P<0.001), natural killer T cells (NKT cells, P<0.0001), type 1 helper T cells (Th1 cells, P<0.0001), and progenitor B cells (pro-B cells, P<0.0001), which play positive roles in immune activation. However, this treatment increased the proportion of regulatory T cells (Tregs, P<0.01), which are inhibitory, and reduced the proportion of epithelial cells, which represents high differentiation and low malignancy (P<0.05, Figure 4A).

In addition, we use the CIBERSORT algorithm to analyze the data from TCIA. The results showed that the expression of PUM1 reduced the proportion of CD8+ T cells (P<0.05), memory B cells (P<0.01) and follicular helper T cells (Tfh cells) (P<0.01), which play a positive role in immune activation, and increased the proportion of resting
PUM1: mRNA expression z-Scores (RNA Seq V2 RSEM)
PUM1: Methylation (HM450)

Spearman: \(-0.29\) (P=8.677e-5)
Pearson: \(-0.23\) (P=2.421e-3)

\[ y = -30.13x + 0.67 \]

\[ R^2 = 0.05 \]

Missense (VUS)
Not mutated
Not profiled for mutations
- Gain
- Diploid
- Shallow Deletion
- Not profiled for CNA
D1

PUM1 - Body-Open_Sea-cg04078732

D2

PUM1 - Body-Open_Sea-cg04976330

D3

PUM1 - Body-Open_Sea-cg08849613

D4

PUM1 - Body-Open_Sea-cg08931023
mast cells, which are inhibitory immune cells (P<0.0001, Figure 4B).

We validated the results in the TIMER database and found that PUM1 expression was significantly positively correlated with the infiltration of memory B cells (r=0.329, P<0.0001), CD8+ T cells (r=0.665, P<0.0001), neutrophils (r=0.468, P<0.0001), macrophages (r=0.517, P<0.0001) and dendritic cells (r=0.503, P<0.0001) but not CD4+ T cells (r=0.106, P=0.159, Figure 4C).

**PUM1 was associated with the IPS, DNA mismatch repair (MMR) and tumor mutation burden (TMB) of pancreatic cancer**

What effect does PUM1 have on the immunotherapy of pancreatic cancer? In this study, we evaluated the relationship between the status of 5 MMR genes and PUM1 expression using TCGA expression profile data. As shown in Figure 5A, PUM1 expression was significantly correlated with MLH1, MSH2, MSH6, PMS2, and EPCAM mutations (P<0.001). TCGA data also showed that patients with low PUM1 expression had a higher TMB than those with high expression (P<0.002, Figure 5B). Furthermore, we thoroughly explored the association between IPS and PUM1 mRNA expression levels in pancreatic cancer patients. In order to estimate the potential of PUM1, the IPS, IPS-PD1/PD2/PD-L1, IPS-CTLA4+ PD1/PD2/PD-L1, and IPS-CTLA4 scores were designed as a biomarker for patients. The following scores were significantly increased in the PUM1 low-expression group compared with the high-expression group (Figure 5C): IPS, P<0.001; IPS-CTLA4, P<0.0001; and IPS-CTLA4+ PD1/PD2/PD-L1, P<0.05. However, there was no significant difference in the IPS of patients treated with PD1/PD2/PD-L1 antibody alone between the high- and low-expression groups (P>0.05, Figure 5C).

These results indicate that the PUM1 low-expression group has few MMR gene mutations, which means that it is more susceptible to somatic mutations and a high TMB. This group also displayed a higher IPS than the high-expression group, which appeared to reflect a more immunogenic phenotype. All the results above indicate that patients with low PUM1 expression benefit more from immunotherapy than those with high PUM1 expression.

**Discussion**

Pancreatic cancer is notorious for its poor prognosis and
Figure 4 Relationship between Infiltration level of immune cells and the mRNA expression level of PUM1 in patients with pancreatic cancer. (A) The gene expression of CLP, NKT cells, Th1 cells, and pro-B cells, inhibitory Tregs, the epithelial cells in PUM1 low-expression and high-expression groups according to xCell database. (B) The gene expression of CD8 T cells, memory B cells, Tfh cells, resting mast cells in PUM1 low-expression and high-expression groups according to TCIA database using CIBERSORT algorithm. (*, P<0.05; **, P<0.01; ***, P<0.001; ****, P<0.0001). (C) The correlation between the expression of PUM1 in PAAD and the immune infiltration level. The PUM1 expression level is significantly positively correlated with infiltrating levels of B cells, CD8 T cells, neutrophils, macrophages, and dendritic cells in PAAD (r=0.329, r=0.665, r=0.468, r=0.317, r=0.503, respectively, P<0.0001), other than CD4 T cells (r=0.106, P=0.159). PUM1, pumilio homologous protein 1; CLP, common lymphoid progenitors; NKT cells, natural killer T cells; Th1 cells, type 1 helper T cells; pro-B cells, progenitor B cells; Tregs, regulatory T cells; Tfh cells, follicular helper T cells; PAAD, pancreatic adenocarcinoma.
Figure 5 IPS and TMB analysis of pancreatic cancer patients. (A) The relationship between IPS and PUM1 mRNA expression level in pancreatic cancer according to TCIA database. The IPS of patients with high and low PUM1 expression level, who were treated with CTLA-4 antibody alone (P<0.0001), CTLA-4 plus PD1/PD2/PDL1 antibody combination therapy (P<0.05) and PD1/PD2/PDL1 antibody alone (P>0.05). (B) The relationship between MMRs and the PUM1 expression level. PUM1 expression was significantly positively correlated with MLH1, MSH2, MSH6, PMS2, EPCAM mutations (P<0.001). (C) The relationship between TMB and the PUM1 mRNA expression level. The TMB of patients with low PUM1 mRNA expression level (n=91) was statistically significant than that with high PUM1 mRNA expression level (n=55, P=0.002). (*, P<0.05; **, P<0.01; ***, P<0.001; ****, P<0.0001). IPS, immunophenoscore; TMB, tumor mutation burden; PUM1, pumilio homologous protein 1; TCIA, The Cancer Imaging Archive.

high mortality. On the one hand, patients with pancreatic cancer usually present with nonspecific symptoms, which makes it difficult to detect early. On the other hand, there are no effective systemic treatments for PAAD, including chemotherapies and molecularly targeted therapies (44,45). This fact has motivated us to carry out further research to explore the pathogenesis of PAAD and to mine new diagnostic and prognostic indicators. Only in these ways can the prognosis of patients with pancreatic cancer be rapidly improved.

PUM1 is a sequence-specific RNA binding protein (24,25) that is involved in various physiological processes, such as the cell cycle, DNA repair, and cell renewal (26-29). In our previous work, we verified that the expression
of PUM1 is correlated with the stage and prognosis of pancreatic cancer (34). In this study, we further explored the potential of PUM1 as a diagnostic and therapeutic target and investigated its underlying mechanisms in pancreatic cancer.

The main tumor markers in the diagnosis of pancreatic cancer at present are CA242 and CA19-9. However, their diagnostic potential is limited because of their restricted sensitivity and specificity (7-11,46-48). In our study, we found that the AUC was 0.9734 when the PUM1 mRNA expression level in tissues was used as a diagnostic marker in pancreatic cancer, and the AUC was close to 1 when of the PUM1 expression level in extracellular vesicles in the blood was used as a diagnostic marker. By contrast, the AUCs of CA19-9 and CA242 were between 0.7 and 0.85 (7). Therefore, PUM1 could serve as a promising diagnostic marker of pancreatic cancer. However, the sample size is small in our analysis; these findings need further validation in a larger database. In addition, our analysis is based on the data of PUM1 mRNA expression and needs to be verified at the protein level.

We also explored the potential regulatory mechanisms of PUM1 expression. Increased methylation of tumor suppressor genes or decreased methylation of oncogenes can promote tumorigenesis (49-55). Large-scale DNA hypomethylation in the genome affects chromosome stability and genome integrity, resulting in an increased chance of cell cancer (50,51,54,56). DNA methylation is a biological process accomplished by DNMTs, which catalyze the covalent addition of a methyl group to the 5-position of cytosine within the CpG island (43). Here, we analyzed the association between PUM1 and DNMT expression levels in a total of 33 tumors. In PAAD, PUM1 expression was significantly positively correlated with the expression of the four methyltransferases (DNMT1, DNMT2, DNMT3A, DNMT3B). This result is consistent with the negative correlation between the PUM1 methylation level and its mRNA expression level according to analysis of the cBioPortal database. In addition, we analyzed 23 methylation sites of PUM1 in pancreatic cancer, and five (cg04976330, cg23281075, cg04078732, cg0891023, cg08849613) were associated with survival. The methylation level of these five sites can be used as a predictor of prognosis in patients with pancreatic cancer.

How does PUM1 play a role in pancreatic cancer? We found that PUM1 was closely related to the immune microenvironment of PAAD. In addition to the oncogenic role of PUM1, the enriched tumor matrix of pancreatic cancer is also a reason for treatment failure. There are abundant non-cancer cell components in pancreatic cancer tissue, which constitute the tumor matrix. The matrix accounts for more than 50% of the total tumor mass. Current studies have shown that the pancreatic tumor microenvironment has a high degree of immunosuppression and inhibits the body’s spontaneous and treatment-induced antitumor immunity (23). The components of the matrix include cellular and chemical components, including fibroblasts, myeloid-derived suppressor cells (MDSCs), tumor-associated macrophages, Tregs and so on. Our study analyzed data from the TCIA and xCell databases and found that Tregs and resting mast cells were more prevalent in the pancreatic cancer tissues of patients with high expression of PUM1 than in those of patients with low expression. These cells are thought to have immunosuppressive effects. Moreover, the number of CD8+ T cells, NKT cells, Tfh cells and memory B cells, which can exert antitumor effects, was small. The suppressive cellular components in the immune microenvironment of pancreatic cancer, on the one hand, form a barrier by producing extracellular matrix, making pancreatic cancer tissue dense and blocking the entry of effector cells and drugs that clear the tumor. On the other hand, the secretion of suppressive cytokines inactivates and depletes killer cells infiltrating the tumor microenvironment so that they cannot kill tumor cells. The infiltration of immune cells in the microenvironment of cancer is an important predictor of prognosis (57,58). Therefore, PUM1 affects the prognosis of patients by suppressing immune cells in the microenvironment.

In addition to the immune microenvironment, PUM1 also has an impact on the efficacy of immunotherapy. Cancer is a genetic disease, which is associated with the accumulation of mutations. Accurate DNA replication and the repair of DNA damage are significant aspects for genome maintenance. DNA MMR was one of the first DNA repair pathways related to cancer predisposition (59). MMR-deficient tumors have the highest mutation rates among all types of cancer types (60). The mutation of DNA MMR genes is associated with a high TMB (61-63). Current studies have shown that a high TMB results in more antigen site exposure, which is more conducive to T cell recognition and tumor cell killing. The higher the TMB is, the better the response to immunotherapy, such as those targeting PD-1/PD-L1, and the greater the overall benefit from immunotherapy (64,65). Using TCGA expression profile data, we found that PUM1 expression was significantly positively correlated with the status of five MMR genes (MLH1, MSH2, MSH6, PMS2, EPCAM mutations).
Consequently, patients with high expression of PUM1 lack MMR genes, which leads to a high TMB. Hence, the IPS, IPS-CTLA4, and IPS-PD1/PD2/PD-L1 + CTLA4 scores were markedly decreased in these patients compared to patients with low expression of PUM1, indicating a worse prognosis. These results suggest that one of the reasons for poor prognosis in patients with high expression of PUM1 is that PUM1 enhances the inhibition of the tumor immune microenvironment in pancreatic cancer.

Overall, in this study, we verified that PUM1 is an oncogene that can inhibit the immune microenvironment and lead to poor prognosis in pancreatic cancer. We provided a novel perspective that PUM1 could be utilized as a diagnostic marker for pancreatic cancer and showed that PUM1 expression was affected by DNA methylation. Online tools, which were on the basis of the most popular bioinformatics theories, were used by us to perform target gene analyses of tumor data from public databases. Compared with traditional chip screening, the advantages of this method include a large sample size, low cost, and simple methods. However, this study still has some limitations. First, all the data were retrieved from public databases, and our findings are required to be verified by external validation. Second, the PAAD samples in the TCGA database are mainly from Caucasians. The etiology and genetic background of PAAD can vary significantly among different ethnic groups. Third, most PAAD samples included in the study were predominantly from early-stage patients who can be treated with surgery. However, in the clinical reality, most PAAD patients are first diagnosed with a disease that is advanced and has an extremely poor prognosis. Data on the level of PUM1 expression in advanced-stage pancreatic cancer are limited. These questions should be addressed in multicenter, large-sample studies, in which clinical samples covering different ethnic groups and PAAD stages should be included.

Acknowledgments

We thank the hospital leaders for their support of our work. We thank Jennifer P. and Courey A. of AJE for English language editing.

Funding: This study was financially supported by National Natural Science Foundation of China (81874211).

Footnote

Reporting Checklist: The authors have completed the MDAR reporting checklist. Available at http://dx.doi.org/10.21037/tcr-20-3295

Conflicts of Interest: All authors have completed the ICMJE uniform disclosure form (available at http://dx.doi.org/10.21037/tcr-20-3295). The authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). The study was approved by Research Ethics Committee of Southwest Hospital, Army Medical University (KY2020138). All the datasets were retrieved from published literature, and written informed consent was confirmed in all studies.

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Cite this article as: Yang Y, Su X, Shen K, Zhang C, Dai H, Ma H, Jiang Y, Shuai L, Liu Z, You J, Min K, Chen Z. PUM1 is upregulated by DNA methylation to suppress antitumor immunity and results in poor prognosis in pancreatic cancer. Transl Cancer Res 2021;10(5):2153-2168. doi: 10.21037/tcr-20-3295