TYPE Original Research PUBLISHED 19 August 2022 DOI 10.3389/fcell.2022.913684

Check for updates

OPEN ACCESS

EDITED BY Jinhui Liu, Nanjing Medical University, China

REVIEWED BY Sihan Chen, First Affiliated Hospital of Anhui Medical University, China Hongjun Yu, Harbin Medical University, China Manish Muhuri, Bioqen Idec, United States

*CORRESPONDENCE Yu Guo, guoyu308@163.com Yangyang Guo, 984054863@qq.com

SPECIALTY SECTION This article was submitted to Cancer Cell Biology, a section of the journal Frontiers in Cell and Developmental Biology

RECEIVED 06 April 2022 ACCEPTED 21 July 2022 PUBLISHED 19 August 2022

CITATION

Hong K, Zhang Y, Yao L, Zhang J, Sheng X, Song L, Guo Y and Guo Y (2022), Pan-cancer analysis of the angiotensin II receptor-associated protein as a prognostic and immunological gene predicting immunotherapy responses in pancancer. Front. Cell Dev. Biol. 10:913684.

doi: 10.3389/fcell.2022.913684

COPYRIGHT

© 2022 Hong, Zhang, Yao, Zhang, Sheng, Song, Guo and Guo. 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.

Pan-cancer analysis of the angiotensin II receptor-associated protein as a prognostic and immunological gene predicting immunotherapy responses in pan-cancer

Kai Hong^{1,2}, Yingjue Zhang³, Lingli Yao^{1,2}, Jiabo Zhang¹, Xianneng Sheng¹, Lihua Song^{1,2}, Yu Guo^{1*} and Yangyang Guo^{1*}

¹Department of Thyroid and Breast Surgery, Ningbo City First Hospital, Ningbo, Zhejiang, China, ²Medicine School, Ningbo University, Ningbo, Zhejiang, China, ³Department of Molecular Pathology, Division of Health Sciences, Graduate School of Medicine, Osaka University, Suita, Osaka, Japan

Background: Understanding interior molecular mechanisms of tumorigenesis and cancer progression contributes to antitumor treatments. The angiotensin II receptor-associated protein (AGTRAP) has been confirmed to be related with metabolic products in metabolic diseases and can drive the progression of hepatocellular carcinoma and colon carcinoma. However, functions of AGTRAP in other kinds of cancers are unclear, and a pan-cancer analysis of AGTRAP has not been carried out.

Methods and materials: We downloaded data from The Cancer Genome Atlas and Genotype-Tissue Expression dataset and The Human Protein Atlas databases and then used R software (version 4.1.1) and several bioinformatic tools to conduct the analysis.

Results: In our study, we evaluated the expression of AGTRAP in cancers, such as high expression in breast cancer, lung adenocarcinoma, and glioma and low expression in kidney chromophobe. Furthermore, our study revealed that high expression of AGTRAP is significantly related with poor prognosis in glioma, liver cancer, kidney chromophobe, and so on. We also explored the putative functional mechanisms of AGTRAP across pan-cancer, such as endoplasmic reticulum pathway, endocytosis pathway, and JAK-STAT signaling pathway. In addition, the connection between AGTRAP and tumor microenvironment, tumor mutation burden, and immune-related genes was proven.

Conclusion: Our study provided comprehensive evidence of the roles of AGTRAP in different kinds of cancers and supported the relationship of AGTRAP and tumorous immunity.

KEYWORDS

AGTRAP, pan-cancer, prognosis, tumor microenvironment, immunity

Introduction

Cancer is one of the major causes of death in humans. Treatments for various types of cancers have become one of the biggest parts of healthcare spending. Publications reported that abnormal expression and mutation of certain genes including methylation-related genes, immune checkpoint genes, immunomodulatory genes, and mismatch repair (MMR) genes could promote the progression of cancers (Chowell et al., 2021; An and Duan, 2022; Bennett et al., 2022; Kraehenbuehl et al., 2022). In addition, in the past several decades, the tumor microenvironment (TME) has been demonstrated to be a vital factor for cancerous development, such as immune cell infiltration or Hashimoto's thyroiditis for cancer progression (Zeng et al., 2018; Gan et al., 2021; Mao et al., 2021). Relaying on abundant online databases, many generelated analyses of cancers were conducted, and a great number of cancer-related genes have been found (Kim et al., 2015; Song et al., 2017; Barneh et al., 2019). Pan-cancer analysis is a valuable way to explore functions and mechanisms of specific genes in various cancers. According to pan-cancer analyses, more detailed research can be designed and conducted.

The renin-angiotensin system (RAS) has been demonstrated to regulate cardiovascular and body fluid homeostasis which causes diseases such as type 2 diabetes mellitus, hypertension, and dyslipidemia (Arendse et al., 2019). Ang II type 1 receptor (AT1R) is an important receptor, and it interacts with the AT1Rassociated protein which is encoded by AGTRAP gene, playing a vital role in activating RAS (Daviet et al., 1999; Lopez-Ilasaca et al., 2003). Furthermore, some studies supported that AGTRAP is also associated with tumor progression. Liu et al. (2021) concluded that AGTRAP expression in hepatocellular carcinoma samples is higher than that in normal tissues and adjacent tissues, which potentially leads to a poor prognosis through regulating nuclear factor-kappa B subunit 1 (NF-KB) and mitogen-activated protein kinase (MAPK) signaling pathways. Sanz-Pamplona et al. (2016) reported that overexpression of AGTRAP was found in colon carcinoma which was related to poor outcome in patients. In addition, AGTRAP-BRAF fusion was validated through qRT-PCR and fluorescence in situ hybridization (FISH) analysis in gastric cancer (Figueiredo et al., 2017). Still, the underlying functions and mechanisms of AGTRAP are unclear. Further explorations are urgently needed to reveal the precise functions and mechanisms of AGTRAP in cancer patients.

TME is made up of tumor, stromal, endothelial, and immune cells (Downs-Canner et al., 2022). According to previous studies, the component immune cells play an important role in controlling cancer progression (Lam et al., 2021). Hence, more and more researchers tend to apply TME to predict prognosis and drug response. DNA MMR is a pathway of correcting DNA strands which are wrongly synthesized (Mas-Ponte and Supek, 2020). Studies presented that a decrease in

MMR activities is related to high microsatellite instability (MSI-H) and high tumor burden (TMB) of tumors (Mandal et al., 2019; Pestana et al., 2022). Simultaneously, in MSI-H tumors, increased tumor-infiltrating lymphocytes (TILs) can be detected (Michael-Robinson et al., 2001). Importantly, researchers have demonstrated how cancers with MMR deficiency benefit from the immune checkpoint blockade can be predicted according to the diversity of MSI (Sahin et al., 2019). In addition, metabolic recombination of tumor cells is a vital method that cancers take to meet their need for growth and their homeostasis in TME (Lunt and Vander Heiden, 2011; Scharping et al., 2021). An and Duan, (2022) suggested that RNA methylation modification plays an important role in metabolic reprogramming which can greatly influence the progression of cancers. These novel markers can be used to predict prognoses of cancer and drug responses and to reveal potential mechanisms of other cancerrelated molecules.

In order to explore the functions and mechanisms of AGTRAP in cancers which were highly unknown, we extracted data from online databases to conduct a pan-cancer analysis. Our study analyzed the expression level of AGTRAP and whether it influenced survival, potential mechanisms of AGTRAP, and correlation between AGTRAP and TME, immune cell infiltration, RNA methylation, MMR, immune checkpoint, immunoregulator, TMB, and neoantigen.

Materials and methods

Gene and protein expression analysis

First, we analyzed the expression difference of AGTRAP (ENSG00000177674) with the detection data of tumor samples and adjacent normal tissues in The Cancer Genome Atlas (TCGA, https://portal.gdc.cancer.gov/) project from Tumor Immune Estimation Resource (TIMER, http://timer.cistrome. org/). TIMER is a resource of immune infiltration, gene expression profile, clinical prognosis, and somatic mutations and copy number alterations for various kinds of cancer (Li et al., 2017). Inputting "AGTRAP" in the "Gene_DE" module, we found that there are some deficiencies in several cancers [e.g., The Cancer Genome Atlas Adrenocortical Carcinoma (TCGA-ACC) and The Cancer Genome Atlas Uterine Serous Carcinoma (TCGA-USC)]. To further demonstrate the abnormal expression of AGTRAP in diverse cancer types, Gene Expression Profiling Interactive Analysis, version 2 tool (GEPIA2, http://gepia2.cancer-pku.cn/#analysis) was used to conduct the analysis for the combination of TCGA and Genotype-Tissue Expression (GTEx) datasets (Tang et al., 2019). TCGA tumor samples were compared to the normal tissues from TCGA and GTEx [*p*-value cutoff = 0.05, log2 fold change (FC) cutoff = 1]. In order to explore the proteomic expression profile of AGTRAP, we conducted the

proteomic expression analysis by UALCAN tool (http://ualcan. path.uab.edu/analysis-prot.html) of the CPTAC (Clinical Proteomic Tumor Analysis Consortium) dataset (Chandrashekar et al., 2017). The keyword "AGTRAP" was input in the "CPTAC analysis" module. The protein expression of AGTRAP in breast cancer, colon cancer, head and neck squamous carcinoma, pancreatic adenocarcinoma, glioblastoma multiforme, and hepatocellular carcinoma was presented, respectively. To discuss AGTRAP expression in different stages of cancer, the "Pathological Stage Plot" module of GEPIA2 was used to visualize the results with violin plots. The log2 [transcripts per million (TPM) +1] was used for log-scale.

Survival analysis

To explore correlations between AGTRAP expression and different kinds of cancer, the log-rank test and a multivariate Cox proportional hazard regression model were performed. The "overall survival (OS)," "disease-specific survival (DSS)," "disease-free survival (DFS)," and "progression-free survival (PFS)" of various cancers were analyzed using the "survival" R package and visualized by forest plots. The hazard ratios (HRs) and 95% confidence intervals (CIs) calculated through Cox analysis were used to identify the survival difference affected by AGTRAP abnormal expression in various types of cancer. The median expression of AGTRAP was used to divide samples into low- and high-expression groups.

Protein–protein interaction network

A protein–protein interaction (PPI) network for AGTRAP was constructed using the STRING website tool (https://stringdb.org/) (Szklarczyk et al., 2019). The medium confidence genes were conserved (interaction score \geq 0.4). Then, correlation analysis of AGTRAP and its related genes in pan-cancer was visualized using the GEPIA2 tool (Tang et al., 2019). We used Pearson's test for the correlation analysis, nonlog scale for calculation, and log-scale axis for visualization. Moreover, the expression of AGTRAP-related genes across pan-caner was visualized by the heatmap using the GEPIA2 tool, and log2 (TPM +1) was used for log-scale (Tang et al., 2019). In addition, OS and DFS of AGTRAP-related genes were also analyzed using the GEPIA2 tool (Tang et al., 2019).

Enrichment analysis

Gene Ontology (GO) and Kyoto Encyclopedia of Genes and Genomes (KEGG) enrichment analyses were conducted to analyze the potential functions of AGTRAP (Kanehisa and Goto, 2000; The Gene Ontology Consortium, 2019). First, we selected the most relevant top 200 genes of AGTRAP through the GEPIA2 tool (Supplementary Material S1). Then, functional annotations and enrichment pathways of AGTRAP and its relevant genes were identified. Moreover, gene set enrichment analysis (GSEA) of single AGTRAP was separately performed in different kinds of cancers (Subramanian et al., 2005). Enrichment analysis was conducted using "limma," "org.Hs.eg.db," "clusterprofiler," and "enrichplot" R packages (Yu et al., 2012).

Immune score and cell infiltration analysis

We downloaded the standardized pan-cancer dataset from the University of California, Santa Cruz (UCSC, https:// xenabrowser.net/) database including TCGA, Therapeutically Applicable Research to Generate Effective Treatments (TARGET), and GTEx cohorts and then extracted the expression data of AGTRAP gene in each sample (Kent et al., 2002). The results of AGTRAP expression were transformed to log(x+0.001), and the gene expression profiles of each kind of tumors were extracted. Afterward, we evaluated the correlation between AGTRAP and immune scores (ImmuneScore, StromaScore, and MicroenvironmentScore) and immune cell infiltration in different kinds of cancers by xCell algorithms using the "IOBR" R package (Zeng et al., 2021). Finally, the significance was identified using Pearson's correlation coefficient by the "psych" R package from the 10,180 samples of 44 types of cancers (Revelle and Revelle, 2015).

Immune checkpoint, RNA modification, and immunoregulator correlation analysis

In order to evaluate the correlation between immune checkpoint, RNA modification, and immunoregulation genes and AGTRAP, Spearman correlation analysis was conducted to analyze the data downloaded from the UCSC database (Myers and Sirois, 2004). Regarding the immune checkpoint, inhibitory and stimulatory genes were analyzed, respectively. In addition, the correlation between "N6-methyladenosine (m(6) A)," "N1-methyladenosine (m(1)A)," and "5-methylcytosine (m(5)C)" modifications and AGTRAP was presented, respectively. Moreover, the relationship between immunoregulation genes and AGTRAP was estimated.

Tumor mutational burden, microsatellite instability, neoantigen, and mismatch repair mutation correlation analysis

The mutation data of each kind of cancer were calculated using the "MAF Tools" R package (Mayakonda et al., 2018). MSI



FIGURE 1

Expression and protein level of AGTRAP in pan-cancer. (A) mRNA expression levels of AGTRAP in different cancers from TCGA data in TIMER. *p < 0.05, **p < 0.01, ***p < 0.01. (B) Analysis of TCGA and GTEx datasets showed differential mRNA expression of AGTRAP in various cancers. *p < 0.05, **p < 0.01, ***p < 0.001. (C) AGTRAP protein expression level in pan-cancer from CPTAC samples. Z-values represented standard deviations from the median across samples for the given cancer types.



The log2 (TPM+1) was used for log-scale.

scores, neoantigen data, and MMR data were obtained from previous studies. MMR genes included: epithelial cell adhesion molecule (EPCAM), PMS1 homolog 2 (PMS2), MutS homolog 6 (MSH6), MutS protein homolog 2 (MSH2), and MutL protein homolog 1 (MLH1) (Latham et al., 2019). Finally, cancers with a number less than 3 were removed from the results. The correlation was analyzed between TMB, MSI scores, neoantigen, and MMR and AGTRAP expression by Spearman's method (Revelle and Revelle, 2015).

Immunohistochemistry

The immunohistochemistry results of AGTRAP pathological expression in diverse cancers were obtained from The Human

Protein Atlas database (THPA, https://www.proteinatlas.org/) (Thul and Lindskog, 2018).

Cell culture and transfection

The breast, pancreatic, and gastric cancer cells (BT-549, PANC-1, and HGC-27) and normal cells (MCF-10A, HPNE, and GES-1) were obtained from the Cell Bank of the Shanghai Institute of Biochemistry and Cell Biology (Shanghai, China). The cell lines were cultivated in Dulbecco's modified Eagle's medium or RPMI-1640, both added with 10% fetal bovine serum (FBS), $100 \,\mu$ g/ml streptomycin, and $100 \,$ U/ml penicillin (Gibco). The cells were infected with si-AGTRAP (GenePharma).



qRT-PCR

All cells were washed with PBS and then homogenized in TRIzol reagent (Invitrogen). Total RNA was extracted and reverse transcribed into the cDNA template, and then SYBR

Green Real-Time PCR Master Mix Plus (Toyobo) was used for qRT-PCR. β -Actin was used as an endogenous reference gene. The primer sequences for amplification were as follows: forward 5'-TGGGGCTGCATTGTATTCTCA-3' and reverse 5'-AGC CACCCAGAAACATGCTTA-3' for AGTRAP and forward



5'-TGACGTGGACATCCGCAAAG-3' and reverse 5'-CTG GAAGGTGGACAGCGAGG-3' for β -actin.

Western blot

To measure the protein expression of AGTRAP and its related signaling pathway, Western blot was conducted on breast cancer, pancreatic cancer, and gastric cancer. Cell proteins were extracted in accordance with manufacturer's instruction. The concentration of proteins was identified using BCA protein assay kit (Beyotime, Shanghai, China). A volume of 35 μ L of proteins from each sample was used for electrophoresis. Membranes were incubated overnight with primary antibodies followed by blocking the membranes with 5% skimmed milk. Subsequently, secondary antibodies (1:5,000) were used for further incubation. Finally, the protein bands were visualized through an enhanced chemiluminescent reagent (Thermo Fisher).

The AGTRAP antibody (F-6) (sc-271367) was purchased from Santa Cruz Biotechnology (Heidelberg, Germany). The mTOR antibody (ab2732) and AKT antibody (ab8805) were purchased from Abcam (Beijing, China). The p-mTOR antibody (CST 5536S) and p-AKT antibody (CST 4060S) were purchased from Cell Signaling Technology (CST, United States). The GAPDH antibody (AP0063) was purchased from Bioworl Technology (United States). The Wilcoxon test was used to compare the data of tumor samples and normal tissues. K–M analysis and Cox regression analysis were used to analyze the survivals. In addition, Spearman's test or Pearson's test was used to conduct the correlation analyses. *p*-values less than 0.05 were considered statistically significant. Annotations were used to show statistical significance: **p* < 0.05, ***p* < 0.01, ****p* < 0.001, and *****p* < 0.0001. R software (version 4.1.1), GraphPad Prism (version 8.4.3), and several bioinformatical tools such as Sangerbox (http://vip.sangerbox.com/home.html) and TIMER were used to conduct the statistical analyses.

Results

Angiotensin II receptor-associated protein expression in pan-cancer

First, AGTRAP mRNA expression in tumor samples and normal tissues was analyzed across TCGA dataset (Figure 1A). AGTRAP expression in tumor samples of BLCA (bladder urothelial carcinoma), BRCA (breast invasive carcinoma), CHOL (cholangiocarcinoma), COAD (colon adenocarcinoma), ESCA (esophageal carcinoma), HNSC (head and neck squamous cell carcinoma), LIHC (liver hepatocellular carcinoma), LUAD (lung adenocarcinoma), READ (rectum adenocarcinoma), STAD (stomach adenocarcinoma), THCA (thyroid carcinoma), UCEC (uterine corpus endometrial carcinoma) (p < 0.001), KIRC (kidney renal clear cell carcinoma), and KIRP (kidney renal papillary cell carcinoma) (p < 0.05) is significantly higher than that in normal tissues. Moreover, in KICH (kidney chromophobe), AGTRAP expression in tumor samples is significantly lower than that in normal tissues (p < p0.001). Notably, HPV-negative HNSC has a higher AGTRAP expression than HPV-positive HNSC.

In view of the deficiency of several kinds of normal tissue data, TCGA and GTEx datasets were pooled together for analyzing. As shown in Figure 1B, significantly higher AGTRAP expression in tumor samples than in normal tissues was presented in GBM (glioblastoma multiforme), LGG (brain lower grade glioma), UCEC, BRCA, CESC (cervical squamous cell carcinoma and endo-cervical adenocarcinoma), ESCA, STES (stomach and esophageal carcinoma), KIRP, COAD, PRAD (prostate adenocarcinoma), STAD, HNSC, LIHC, SKCM (skin cutaneous melanoma), BLCA, THCA, READ, OV (ovarian cystadenocarcinoma), PAAD (pancreatic serous adenocarcinoma), TGCT (testicular germ cell tumors), UCS (uterine carcinosarcoma), CHOL (p < 0.0001), LUAD (p < 0.001), and KIPAN (pan-kidney) (p < 0.01). Significantly lower AGTRAP expression in tumor samples than in normal tissues was demonstrated in LUSC (lung squamous cell

carcinoma), ALL (acute lymphoblastic leukemia), LAML (acute myeloid leukemia), KICH (p < 0.0001), and PCPG (pheochromocytoma and paraganglioma) (p < 0.05).

Abnormal levels of cancer-related proteins are strongly related to cancer progression. Thus, in addition to the mRNA expression analysis, AGTRAP data from CPTAC was analyzed by comparing the tumor samples to normal tissues. Finally, significant high expression of AGTRAP was found in breast cancer, colon cancer, head and neck squamous carcinoma, pancreatic adenocarcinoma, glioblastoma multiforme, and hepatocellular carcinoma (Figure 1C).

In order to evaluate the expression of AGTRAP in different pathological stages in pan-cancer, we conducted the "Pathological Stage Plot" module in GEPIA2. A significant difference of AGTRAP expression in different pathological stages was demonstrated across various cancer types, such as ESCA, KICH, BLCA, KIRC, LIHC, PAAD, and UCS (p < 0.05) (Figure 2).

Survival analysis of abnormal angiotensin II receptor-associated protein expression in pan-cancer

TCGA cancer cases were divided into low- and highexpression subsets according to AGTRAP expression, and then the correlation between AGTRAP expression and prognosis was analyzed. Highly expressed AGTRAP was associated to significantly poor OS in LGG, LIHC, and UVM (uveal melanoma) (p < 0.05) (Figure 3A), significantly poor DFS in LGG and STAD (p < 0.05) (Figure 3B), significantly poor DFS in KIRC, LGG, LIHC, STAD, and UVM (Figure 4A), and significantly poor PFS in LGG, HNSC, LIHC, and STAD (p < 0.05) (Figure 4B). Contrarily, highly expressed AGTRAP showed better OS in MESO (mesothelioma) and THCA (p < 0.05) (Figure 3A), better DFS in COAD (p < 0.05) (Figure 3B), better DSS in BRCA and THCA (p < 0.05) (Figure 4A), and better PFS in BRCA (p < 0.05) (Figure 4B).

Protein-protein interaction network and correlation and survival analyses of angiotensin II receptor-associated protein-related genes

The top 10 related genes of AGTRAP were identified through STRING, namely, AGTR1, JAK2, RAF1, CAMK2A, CAMK2B, CAMK2D, CAMK2G, CLCN6, ARAF, and PITPNC1 (Figure 5A). As shown in Figure 5B, the expression of AGTRAP was positively correlated with AGTR1 and ARAF. A significantly negative correlation was confirmed between AGTRAP and CAMK2A, CAMK2B, CAMK2D, CAMK2G, CLCN6, JAK2, and PITPNC1. In addition, difference in the



Analysis of top 10 AGTRAP-related genes. (A) AGTRAP interaction network. (B) Expression correlation between AGTRAP and AGTRAP-related genes (AGTR1, JAK2, RAF1, CAMK2A, CAMK2B, CAMK2D, CAMK2G, CLCN6, ARAF, and PITPNC1). (C) Expression of AGTRAP-related genes in different cancer samples and normal tissues. (D) Overall survival (OS) analysis of AGTRAP-related genes in pan-cancer. (E) Disease-free survival (DFS) analysis of AGTRAP-related genes in pan-cancer.



expression of AGTRAP-related genes was found in various cancers in TCGA pan-cancer datasets (Figure 5C). A significant difference in OS and RFS of the AGTRAP-related genes was presented in the heatmaps. Notably, the same gene shows contrary affects in different cancers. Survival analysis

showed that the high expression of AGTRAP-related genes mostly leads to poor prognoses in ACC, BLCA, KICH, LGG, LIHC, LUSC, MESO, SARC (sarcoma), and UVM and leads to better outcomes in CHOL, KIRC, KIRP, and PAAD (Figures 5D,E). In addition, to better represent the expression of



AGTRAP-related genes in pan-cancer, we conducted the analysis in the TIMER database. The results showed that the AGTRAPrelated genes have significantly abnormal expression in diverse cancer types. (Figure 6).

Angiotensin II receptor-associated protein and related gene functional enrichment analysis

To further explore the molecular mechanisms of AGTRAP in pan-cancer, enrichment analysis was conducted using the data of AGTRAP and its top 200 related genes. As shown in Figure 7A, AGTRAP is mostly involved in "protein processing in the endoplasmic reticulum pathway," "alcoholism pathway,"

"endocytosis pathway," "prostate cancer pathway," and "amino sugar nucleotide sugar metabolism pathway." In addition, AGTRAP is mostly enriched in hallmarks such as "androgen response," "glycolysis," "protein secretion," "peroxisome," and "DNA repair." Regarding the GO enrichment analysis, "intracellular transport," "intracellular protein transport," "cellular component disassembly," and "golgi vesicle transport" are the most relative functional annotations of biological process; "endoplasmic reticulum," "organelle subcompartment," "golgi apparatus," and "nuclear outer membrane endoplasmic reticulum membrane network" are mostly related to cellular component; "transporter activity," "cell adhesion molecule binding," "cadherin binding," and "cation transmembrane transport activity" are significantly associated to molecular function (Figure 7B).



Afterward, enrichment pathways of single AGTRAP in different kinds of cancers were evaluated to further reveal the potential mechanism (Figure 8). From GSEA, enrichment of "olfactory transduction pathway" was found in most types of cancers including DLBC, ESCA, BLCA, BRCA, CESC, COAD, GBM, and KIRC. In addition, the "neuroactive ligand-receptor interaction pathway" was enriched in BLCA, BRCA, CHOL, GBM, HNSC, and KIRC. Notably, the "autoimmune thyroid disease pathway," "cytosolic DNA sensing pathway," "olfactory transduction pathway," "RIG-I-like receptor signaling pathway," "neuroactive ligand-receptor interaction pathway," and "taste transduction pathway" were enriched in DLBC, BRCA, and COAD, which means that AGTRAP might have some similar functions in these three kinds of cancers. Importantly, AGTRAP also enriched in some immune-related pathways including "chemokine signaling pathway," "cytokine-receptor interaction pathway," "natural killer cell-mediated cytotoxicity pathway," "autoimmune thyroid disease pathway," "cytosolic DNA sensing pathway," "regulation of autophagy pathway," "RIG-I-like receptor signaling pathway," "antigen processing and presentation pathway," "JAK-STAT signaling pathway," "toll-like receptor "neuroactive ligand-receptor signaling pathway," and interaction pathway" and some metabolism-related pathways including "taurine and hypotaurine metabolism pathway," "drug metabolism cytochrome P450 pathway," "metabolism of xenobiotics by the cytochrome P450 pathway," "retinol metabolism pathway," "glycine, serine, and threonine metabolism pathway," and "arachidonic acid metabolism pathway," which confirmed that AGTRAP is a gene whose functions are associated with immunity and metabolism.

Tumor microenvironment correlation analysis

As shown in Figure 9, a significantly positive association was found between AGTRAP and ImmuneScore in LGG, NB (neuroblastoma), BLCA, SARC, UVM, PCPG, KIPAN, LIHC, KICH, LAML, ALL, and GBM; between AGTRAP and StromalScore in LGG, NB, PCPG, COAD, LUSC, and THYM; and between AGTRAP and MicroenvironmentScore in LGG, NB, BLCA, SARC, UVM, PCPG, LAML, ALL, and GBM. Moreover, a significantly negative correlation was confirmed between AGTRAP and ImmuneScore in THCA, BRCA, THYM, OV, STES, and STAD; between AGTRAP and StromalScore in PRAD, KIPAN, LIHC, TGCT, CESC, MESO, OV, STES, and HNSC; and between AGTRAP and MicroenvironmentScore in PRAD, THCA, THYM, OV, STES, HNSC, and STAD.

Then, immune cell infiltration analysis was conducted to further expose the association between AGTRAP expression and tumor microenvironment in pan-cancer (Figure 9). The heatmap shows that in almost all kinds of cancers, AGTRAP expression is



variously related with immune and stromal cells. Notably, LGG, NB, BLCA, SARC, UVM, and PCPG are cancers which have more positive coexpression of AGTRAP and cell infiltration, while OV, THYM, LAML, UCEC, and STES have more negative coexpression. In addition, M2 macrophages, mesangial cells, monocytes, and epithelial cells show positive coexpression in most kinds of cancers and CD4⁺ T cells including memory and naïve cells, myocytes, naïve B cells, neutrophils, pericytes, plasma cells, and platelets show negative coexpression in most cancers.

Immune checkpoint, RNA methylation, and immunomodulator correlation analysis

The correlation of AGTRAP expression and 60 immune checkpoint genes including inhibitory and stimulatory genes was analyzed (Figure 10). AGTRAP expression in different kinds of cancer is variously associated with immune checkpoint genes. In NB, SARC, UVM, DLBC, GBM, LGG, TGCT, KIRC, COAD, PCPG, SKCM, KIPAN, BLCA, CHOL, THYM, LIHC, PAAD, ACC, and OV, AGTRAP expression was more strongly related with immune checkpoint genes than others. In addition, CD276, TGFB1, C10orf54, IL10, CD274, BTN3A1, BTN3A2, CXCL10, CCL5, GZMA, PRF1, CD40, TNFRSF18, TNFRSF4, TNFRSF14, TNFRSF9, ITGB2, ICAM1, HMGB1, CX3CL1, CD70, ICOS, CD27, CD40LG, ENTPD1, IL1A, CD80, CXCL9, IFNG, TNF, IL1B, and IL2RA were more significantly associated with AGTRAP. From the result, it can obviously be concluded that AGTRAP is more related with the stimulatory immune checkpoint.

Regarding the RNA modification genes, correlation analysis was performed for these genes and AGTRAP across various cancers. As shown in Figure 11, AGTRAP expression in ALL, NB, SARC, HNSC, LUSC, LGG, OV, ACC, KICH, TGCT, BLCA, GBM, CHOL, LIHC, PAAD, ESCA, KIPAN, KIRC, UVM, COAD, WT (Wilms' tumor), THCA, STAD, and STES relates with more modification genes. Moreover, higher correlation modification



genes include: m(1)A writer: TRMT61A; m(1)A readers: TYHDF1 and TYHDF2; m(1)A eraser: ALKBH3; m(5)C writers: NSUN4 and NSUN5; m(5)C reader: ALYREF; m(5)C eraser: TET2; m(6)A writers: KIAA1429; m(6)A readers HNRNPA2B1, HNRNPC, YTHDF1, YTHDF2, and ELAVL1; and m(6)A eraser: ALKBH5.

Figure 12 shows that the immunoregulatory genes of CESC, LUSC, SKCM, COAD, THYM, STES, NB, ACC, PCPG, KIPAN, KIRC, DLBC, UVM, GBM, LGG, OV, LIHC, PAAD, BLCA, and SARC were highly related with AGTRAP. Importantly, MHC-

related genes were more associated with AGTRAP expression than the chemokine-related and receptor-related genes in pan-cancer.

Tumor mutational burden, microsatellite instability, neoantigen, and mismatch repair mutation correlation analysis

TMB, MSI, and neoantigen are the three novel biomarkers related with immunotherapy responses. Correlation analyses



were performed between these biomarkers and AGTRAP expression in pan-cancer. The AGTRAP expression level is significantly associated with TMB in CHOL (R < -0.2), ACC, and KICH (R > 0.2) (Figure 13A). Moreover, a significant correlation between AGTRAP and MSI is found in KIPAN and USC (R < -0.2), and GBM (R > 0.2) (Figure 13B). In addition, the relationship between AGTRAP and neoantigen was estimated, and CHOL and TGCT showed a negative correlation (R < -0.2), while DLBC and COAD showed a positive correlation (R > 0.2) (Figure 13C).

MMR genes play a vital role in maintaining gene stability, such as correcting base mismatches and deletions and insertion errors (Liu et al., 2017). Therefore, coexpression analysis of MMR genes and AGTRAP in pan-cancer was conducted and presented by the heatmap. Figure 13D shows that all five MMR genes we chose are significantly related with various cancers, in which PMS2, MSH6, and MSH2 are more related with AGTRAP than the other MMR genes. In addition, GBM, HNSC, LUSC, PRAD, THCA, and THYM are cancers in which AGTRAP expression is significantly related with all the five MMR genes.

Immunohistochemistry of the angiotensin II receptor-associated protein in pancancer

Immunohistochemistry analysis highlighted the expression of AGTRAP in breast cancer, cervix cancer, colorectal cancer, endometrial cancer, glioma, head and neck cancer, liver cancer, lung cancer, lymphoma, melanoma, ovarian cancer, pancreatic cancer, prostate cancer, renal cancer, skin cancer, stomach cancer, testis cancer, thyroid cancer, and urothelial cancer (Figure 14).



qRT-PCR of the angiotensin II receptorassociated protein in pan-cancer

To further verify the abnormal expression of AGTRAP in pan-cancer, we conducted qRT-PCR to detect the mRNA expression in breast cancer, pancreatic cancer, and gastric cancer. The results showed that AGTRAP is significantly highly expressed in breast cancer, pancreatic cancer, and gastric cancer, which verifies our bioinformatic analysis (Supplementary Figure S1).

Western blot of the angiotensin II receptor-associated protein and the signaling pathway in pan-cancer

In addition, we performed Western blot in pan-cancer. Subsequently, high expression of AGTRAP was identified in

breast cancer, pancreatic cancer, and gastric cancer (Supplementary Figure S2A). To explore the mechanism underlying AGTRAP, the protein expression of AKT/mTOR axis was measured by Western blot. The results showed that p-AKT and p-mTOR expression of si-AGTRAP was significantly lower than the expression of control and si-NC (Supplementary Figure S2B).

Discussion

As a protein that regulates visceral obesity-related metabolism, AGTRAP has been widely researched in metabolic disorders (Maeda et al., 2013). Accumulating evidence confirmed that AT1R can regulate the RAS signaling pathway, thereby affecting the handling of renal sodium (Tamura et al., 2022). The overexpression of AT1R would excessively activate the RAS signaling, leading to the



retention of sodium which is related to hypertension (Arendse et al., 2019). As a binding protein of the carboxyl-terminal domain, the AT1R-associated protein can suppress the hyperactivation of AT1R (Tamura et al., 2013). Previous studies confirmed that AT1R-associated proteins can inhibit the angiotensin-dependent hypertension by decreasing the sodium reabsorption (Ohsawa et al., 2014). In addition to metabolic functions, several bioinformatic analyses demonstrated that AGTRAP can be used as a component of prognostic signatures in brain lower grade glioma, tongue squamous cell carcinomas, and melanoma, revealing the potential of prognostic ability of AGTRAP (Loftus et al., 2017; Zeng et al., 2019; Xiong et al., 2020). However, details of AGTRAP about expression, survival, mechanism, and treatment response in diverse cancers are highly unclear.

In our study, the high expression of AGTRAP was found in 14 kinds of cancer and low expression in 1 cancer type from TCGA datasets. Moreover, the analysis of combination of TCGA and GTEx datasets suggested that AGTRAP was highly expressed in 26 kinds of cancer and lowly expressed in six cancer types. In addition, the abnormal protein expression of AGTRAP has been confirmed in the CPTAC samples across breast cancer, colon cancer, head and neck cancer, pancreatic cancer, glioblastoma multiforme, and hepatocellular carcinoma. Survival analysis showed that 12 kinds of cancers have poorer OS significantly related with the overexpression of AGTRAP and 2 types of cancers have better OS associated with overexpressed AGTRAP. According to our results, we found that our study is consistent with the past studies which researched AGTRAP in glioma and



Immunohistochemistry of AGTRAP expression (normal tissue is on the left and tumor tissue is on the right) (A) Breast cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/breast-cancer). (B) Cervix cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/colecatel-cancer). (C) Colorectal cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/colecatel-cancer). (C) Colorectal cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/endometrial+cancer). (E) Glioma (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/leided+and+neck+cancer). (G) Renal cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/leided+and+neck+cancer). (G) Renal cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/leided+and+neck+cancer). (G) Renal cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/skin+cancer). (I) Stomach cancer (https://www.proteinatlas.org/ENSG0000177674-AGTRAP/pathology/liver+cancer). (K) Lung cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/liver+cancer). (K) Lung cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/ lymphoma). (M) Ovarian cancer (https://www.proteinatlas.org/ENSG0000177674-AGTRAP/pathology/ lymphoma). (P) Testis cancer (https://www.proteinatlas.org/ENSG0000177674-AGTRAP/pathology/patceatic+cancer). (Q) Prostate cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/patceatic+cancer). (Q) Thyroid cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pathology/patceatic+cancer). (Q) Thyroid cancer (https://www.proteinatlas.org/ENSG0000177674-AGTRAP/pathology/patceatic+cancer). (Q) Thyroid cancer (https://www.proteinatlas.org/ENSG00000177674-AGTRAP/pat

hepatocellular carcinoma (Xiong et al., 2020; Liu et al., 2021). However, in our study, the high expression of AGTRAP in colon cancer showed better survival which is contrary to a previous study (Sanz-Pamplona et al., 2016). In view of the deficiency of studies of AGTRAP in cancers, relevant research studies are urgently needed.

TABLE 1 AGTRAP-related genes.

Gene	Official full name	Ensemble	Function
AGTR1	Angiotensin II receptor type 1	ENSG00000144891	AGTR1 inhibits the progression of lung adenocarcinoma and promotes the proliferation of ovarian cancer
JAK2	Janus kinase 2	ENSG00000096968	JAK2 promotes the development and metastasis of lung adenocarcinoma and enhances osteosarcoma growth
RAF1	Raf-1 proto-oncogene	ENSG00000132155	RAF1 promotes progression and predicts poor survival of liver cancer; RAF1 amplification drives bladder tumorigenesis through activating the MAPK pathway
PITPNC1	Phosphatidylinositol transfer protein cytoplasmic 1	ENSG00000154217	PITPNC1 is overexpressed in metastatic breast, melanoma, and colon cancers and enhances vesicular secretion capacity in malignancy; PITPNC1-mediated fatty acid metabolic reprogramming promotes gastric cancer metastasis
ARAF	A-raf proto-oncogene	ENSG00000078061	ARAF promotes gall bladder tumorigenesis; ARAF mutation increases resistance of belvarafenib in melanoma
CLCN6	Chloride voltage-gated channel 6	ENSG00000011021	CLCN6 mutation causes West syndrome
CAMK2A	Calcium/calmodulin-dependent protein kinase II alpha	ENSG00000070808	CAMK2A promotes triple negative breast cancer metastasis, facilitates lung adenocarcinoma progression, and leads to poor prognosis
CAMK2B	Calcium/calmodulin-dependent protein kinase II beta	ENSG00000058404	CAMK2B inhibits papillary renal cell carcinoma and breast cancer proliferation
CAMK2D	Calcium/calmodulin-dependent protein kinase II delta	ENSG00000145349	CAMK2D increases resistance to cisplatin in ovarian cancer, suppresses gastric cancer progression and metastasis, and inhibits the growth of liver cancer
CAMK2G	Calcium/calmodulin-dependent protein kinase II gamma	ENSG00000148660	CAMK2G drives cisplatin resistance in ovarian cancer, promotes breast cancer progression and metastasis, and enhances lung tumorigenesis

The PPI network revealed the most related 10 genes of AGTRAP, namely, AGTR1, JAK2, RAF1, CAMK2A, CAMK2B, CAMK2D, CAMK2G, CLCN6, ARAF, and PITPNC1. As shown in Table 1, most of them have been confirmed to influence tumor progression in various cancers, despite some genes showing contrary functions in different cancers. Based on our KEGG analysis, AGTRAP is related with protein processing in ER also called unfolded protein response (UPR) or endoplasmic reticulum (ER) stress (Hetz and Saxena, 2017; Hetz et al., 2020). ER stress can degrade misfolded proteins, decrease protein translation, and induce apoptosis (Wang et al., 2017; You et al., 2021). Researchers have reported that CHOP, Bcl-2 family, caspase-12, and JNK were components of ER stress related to the apoptotic pathway, whose cancer-related mechanisms have been demonstrated in various cancers (Ashkenazi et al., 2017; Kim et al., 2018; Tan et al., 2006). Therefore, there might be some potential mechanisms between these molecules and AGTRAP in cancers that need to be further explored. Furthermore, alcoholism is another related pathway of AGTRAP in cancer progression. From past studies, associations between alcoholism and various kinds of cancer have been widely explored. Lin et al. (2020) found that longer alcoholism history is related with the higher risk of colorectal cancer. In the study of Sun et al. (2020), an anti-alcoholism drug disulfiram showed the effect of causing immunogenic cell death in radiation-resistant breast cancer stem cells. Interestingly, Zhang et al. (2019) supported that disulfiram can induce autophagy-dependent apoptosis of pancreatic and

breast cancer cells by evoking ER stress, which is accomplished through activation of IRE1a. Therefore, there might be some synergetic effects between ER stress and alcoholism, through which AGTRAP plays its role in cancerous development. Moreover, our analysis showed the relationship between AGTRAP and endocytosis. As a method of capturing materials from the extracellular environment, there takes place a special endocytic process named macropinocytosis, which can degrade protein to supply amino acids for cancer cell development and remove damaged material from the plasma membrane to restore membrane integrity (Commisso et al., 2013). Notably, it is this function of catching materials that makes it possible to mediate targeting therapy (Xiao et al., 2021). Nanomedicines (NMs) can effectively increase therapeutic efficacy and decrease toxicity for cancer therapy. Through different endocytosis processes, NMs can enter cells (Patel et al., 2019; Wang et al., 2020). If it is possible that drugs can selectively enter into cancer cells instead of delivering into healthy cells, cancer treatment would be highly efficient with low toxicity. In light of the KEGG analysis, AGTRAP may have potential mechanisms related with endocytosis in cancers, perhaps mediating the endocytosis process, which means that it is likely to use AGTRAP to reveal selectivity of drugs in cancer cells, which is worthy to be explored. In addition, AGTRAP is associated with the prostate cancer pathway according to enrichment analysis, which is consistent with our expression analysis that AGTRAP is highly expressed in prostate cancer. We also found the correlation between AGTRAP and androgen response, and

since prostate cancer is strongly related with androgen, we believe that our analysis revealed the possible connection between AGTRAP and prostate cancer. According to the GO analysis, the most related biological process of AGTRAP is intracellular transport, which plays a role in direct movement of materials within cells (Clark, 2020). In addition, as the highest related molecular function of AGTRAP, transporter activity including lipid transporter activity, protein transporter activity, and transmembrane transporter activity have the function of enabling the directed movement of substances (Fuji et al., 2009; Nishimura and Stefan, 2020). Furthermore, combining the cellular component result, in which ER is the most enriched annotation, we inferred that AGTRAP is possible to influence progression through ER-related cancer material transportation. In our GSEA, we demonstrated that AGTRAP enriched in many immune-related pathways: "natural killer cell-mediated cytotoxicity pathway," "chemokine signaling pathway," "RIG-I-like receptor signaling pathway," "toll-like receptor signaling pathway," and "JAK-STAT signaling pathway"; and some metabolismrelated pathways: "taurine and hypotaurine metabolism pathway" and "glycine, serine, and threonine metabolism pathway." Evidence has confirmed that natural killer (NK) cells play a vital role in immunosurveillance and immunoediting, correlating with other immunity and assisting immunotherapy (Malmberg et al., 2017; Pugh-Toole et al., 2022). Veneziani et al. (2022) found that the function of NK cells can be strengthened by toll-like receptor 8 agonists, meaning that the interaction between NK cells and toll-like receptor signaling pathway, both of which are potential pathways that AGTRAP acts on. In addition, the enrichment of the chemokine signaling pathway was consisted in the abundant correlation between immunomodulators and AGTRAP, meaning that AGTRAP has functions in regulating immunomodulators, promoting tumor progression. In addition, RIG-I-like receptor and tolllike receptor are two nucleic acid sensors whose actions for innate immunity and potential for immunotherapy have been identified (Kanzler et al., 2007; Loo and Gale, 2011). Furthermore, previous studies proved that taurine, serine-glycine, and threonine metabolism could promote tumorigenesis and cancer cell proliferation in various kinds of cancers through diverse pathways (Locasale, 2013; Kim et al., 2014; Lin et al., 2018). In conclusion, our study indicated that AGTRAP might regulate these metabolismrelated and immune-related pathways, leading to tumor development.

A total of three novel markers used to predict immunotherapy-responses including TMB, MSI, and neoantigen were evaluated in pan-cancer. Immune checkpoint inhibitors (ICIs) have been demonstrated to be efficient in treating various kinds of cancers (Darvin et al., 2018). Currently, despite having found a lot of biomarkers for ICI and OS such as different immune checkpoints, PD-L1 is the only one which has been widely confirmed, such as the validated effectiveness of pembrolizumab for nonsmall cell lung cancer selected with PD-L1 > 50% and the combination of atezolizumab and nab-paclitaxel for metastatic triple negative breast cancer with PD-L1 > 1% (Schmid et al., 2018; Reck et al., 2019). Nevertheless, the prediction ability of PD-L1 is imperfect because of its heterogeneity and lability (Kim and Chung, 2019). Therefore, TMB, which is defined as the number of nonsynonymous exonic mutations per megabase, is used to predict the response of ICI (Fumet et al., 2020). Another biomarker, MSI-H, is demonstrated to have better prognoses in colon cancer patients (Latham et al., 2019). Moreover, neoantigens as mutant peptides are produced through antitumor immune response, which is elicited by somatic alterations where mutated peptide fragments are generated and then presented on the class I major histocompatibility complex (MHC-1) (Yarchoan et al., 2017). Since neoantigens can impel anticancer immunity and is tumor-specific, it has been regarded as potential therapeutic targets (Schumacher and Schreiber, 2015). Through our correlation analysis, the overexpression of AGTRAP predicts better immunotherapy responses in ACC, KICH, GBM, DLBC, and COAD. In view of the prediction ability of TMB, MSI, and neoantigen in diverse cancers, potential of AGTRAP for predicting drug responses combined with these novel biomarkers in pan-cancer is promising. Immune cells play important roles in tumorigenesis. In cancerous progression, immune cells can effectively eliminate tumor cells in early stages, but cancer cells gradually generate the tumor immune escape and influence the state of immune cells (Bremnes et al., 2011; Gonzalez et al., 2018). Among immune cells, neutrophils, natural killer cells, macrophages, dendritic cells, B cells, and T cells are those which are mostly related with cancer development. From our results, we found that correlations between AGTRAP expression and diverse immune cells can be found in almost all kinds of cancers. In addition, AGTRAP expression is significantly related with ImmuneScore in 19 types of cancers and with StromalScore in 17 types of cancers, which also reveals the relationship between AGTRAP and cancer immunity. Importantly, we found that expression positively AGTRAP was related with M2 macrophage infiltration among 25 cancer types. M2 macrophage is a kind of tumor-associated macrophages (TAMs) that play a vital role in forming TME (Pan et al., 2020). Previous research confirmed that M2 macrophage has functions of promoting angiogenesis, reconstructing tissue, repairing injury, and facilitating tumorigenesis and progression (Zhu et al., 2017; Annamalai et al., 2018). Chen et al. (2017) demonstrated that increased M2 macrophage was associated with poor prognosis and metastasis in breast cancer. Also, Yin et al. (2016) uncovered that M2 macrophage could

promote spheroid formation and tumor growth stimulating metastasis in ovarian cancer. By combining previous studies and our analysis, we could speculate that AGTRAP might increase the M2 macrophage infiltration in TME which leads to tumor development. All of the factors we discussed previously have been used to evaluate the TME. As the biomarker for cancers, TME can be used to predict the responses to treatment and prognoses of patients. Based on our correlation analyses, AGTRAP shows the relationship with TME that makes the precise prediction for drug responses and outcomes possible.

In order to further reveal the mechanism of AGTRAP in influencing cancer progression, we analyzed the correlation between AGTRAP and some vital genes including immune checkpoint, RNA methylation, immunoregulator, and MMR. Studies have demonstrated that MMR plays vital roles in maintaining stability and integrity of the genome in cells (Li, 2008). Abnormality of MMR can augment the mutation frequency of related genes which would promote cancer growth and affect therapy responses (Green et al., 2011). The chemokine receptor, MHC, has been widely confirmed to increase tumorous growth and metastasis (Zlotnik et al., 2011; Nagarsheth et al., 2017; Miles et al., 2021). Bruchard et al. (2022) found that CCL20 can boost ILC3-dependent anticancer immunity and increase cancer sensitivity to immunotherapy. Dai et al. (2021) demonstrated the efficacy of CXCL13+CD8+T-cell infiltration for predicting the prognosis and being an immunotherapeutic target in clear cell renal cell carcinoma. RNA methylation, a common chemical modification in eukaryotic RNAs, is related with tumorous development acting through being catalyzed by writers, being removed by erasers, and interacting with readers (Lan et al., 2019). For instance, METTL14, a writer of m(6)A methylation, demonstrated downregulation in gastric cancer tissues and poor prognosis in gastric cancer patients (Fan et al., 2022). In addition, TRMT6/TRMT61A-mediated m(1)A methylation can drive liver tumorigenesis through cholesterol metabolism (Wang et al., 2021) and NSUN2-mediated and YBX1-mediated m(5)C methylation confirmed driving bladder carcinoma tumorigenesis (Chen et al., 2019). PD-l/PD-L1 is the most common immune checkpoint and of great significance for cancer, but others should not be ignored. Analyses of cancer-related genes combining with immune checkpoints have huge potentials for researching tumorigenesis and treatments. According to our study, AGTRAP shows great relationships with immune checkpoint, RNA methylation, immunoregulator, and MMR genes in pancancer, which means that there possibly exist some interactions between them influencing cancer progression. However, this field is almost blank and needs further exploration.

To further verify the protein expression of AGTRAP in pancancer, Western blot was also conducted on breast cancer, pancreatic cancer, and gastric cancer. Our results demonstrated that AGTRAP was highly expressed in these cancer types, which is consistent with the analysis across CPTAC dataset. Subsequently, the AKT/mTOR axis was detected both in breast cancer and gastric cancer. The results revealed that p-AKT and p-mTOR protein expression was significantly highly expressed, meaning that the AKT/mTOR axis was the possible downstream of AGTRAP. Previous studies reported that AKT and mTOR proteins function by phosphorylation (Wang et al., 2022; Zhu et al., 2022), and AGTRAP can promote phosphorylation of AKT and mTOR, initiating the AKT/mTOR axis. Accumulating evidence has demonstrated the functions of p-AKT and p-mTOR in promoting the progression of breast cancer and gastric cancer (Xu et al., 2021; Khan et al., 2022). Our study reveals that a novel biomarker, AGTRAP, can regulate the AKT/mTOR axis.

Nevertheless, the study still has some limitations. First, although with a global analysis for pan-cancer, details about each cancer type are insufficient. Thus, we will conduct deep analysis for specific cancer in our future studies. In addition, as a pan-cancer bioinformatic study, although with reasonable logic and comprehensive analysis for diverse cancers, combining with the verification of AGTRAP expression in three cancer types, deeper mechanisms of AGTRAP in each cancer type still need to be explored in the future.

In conclusion, our analysis is the first to explore the expression difference of AGTRAP between pan-cancer samples and their normal tissues, analyzing the prognostic impact and potential mechanisms of AGTRAP and its related genes in different kinds of cancers and revealing the correlation between AGTRAP and TME, TMB, MSI, neoantigen, and immune-related genes, which can assist predicting immunotherapy responses. In this study, we showed that AGTRAP and its related genes show abnormal expressions in different types of cancers that might affect tumor progression through regulating immune-related and metabolism-related pathways. Moreover, the abnormal expression of AGTRAP is associated with several immune-related biomarkers and could effectively predict drug responses, meaning that AGTRAP has the potential to become a target for antitumor immunotherapy.

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.

Author contributions

KH, YZ, and YYG designed the study and drafted the manuscript. KH, YZ, LY, JZ, and YYG wrote the manuscript. JZ, XS, LS, and YG searched the publications and collected the

data. KH and YYG analyzed the data. The final manuscript was reviewed by all authors and approved for publication.

Acknowledgments

The authors thank all the members for contributing to this 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

An, Y., and Duan, H. (2022). The role of m6A RNA methylation in cancer metabolism. *Mol. Cancer* 21 (1), 14. doi:10.1186/s12943-022-01500-4

Annamalai, R. T., Turner, P. A., Carson, W. F., Levi, B., Kunkel, S., and Stegemann, J. P. (2018). Harnessing macrophage-mediated degradation of gelatin microspheres for spatiotemporal control of BMP2 release. *Biomaterials* 161, 216–227. doi:10.1016/j.biomaterials.2018.01.040

Arendse, L. B., Danser, A. H. J., Poglitsch, M., Touyz, R. M., Burnett, J. C., Llorens-Cortes, C., et al. (2019). Novel therapeutic approaches targeting the reninangiotensin system and associated peptides in hypertension and heart failure. *Pharmacol. Rev.* 71 (4), 539–570. doi:10.1124/pr.118.017129

Ashkenazi, A., Fairbrother, W. J., Leverson, J. D., and Souers, A. J. (2017). From basic apoptosis discoveries to advanced selective BCL-2 family inhibitors. *Nat. Rev. Drug Discov.* 16 (4), 273–284. doi:10.1038/nrd.2016.253

Barneh, F., Mirzaie, M., Nickchi, P., Tan, T. Z., Thiery, J. P., Piran, M., et al. (2019). Integrated use of bioinformatic resources reveals that co-targeting of histone deacetylases, IKBK and SRC inhibits epithelial-mesenchymal transition in cancer. *Brief. Bioinform* 20 (2), 717–731. doi:10.1093/bib/bby030

Bennett, J. A., Fleming, G. F., Kurnit, K. C., Mills, K. A., and vanWeelden, W. J. (2022). Tumor board presentation of a woman with metastatic, hormone receptorpositive, mismatch repair-deficient endometrial cancer. *CA Cancer J. Clin.* 72, 102. doi:10.3322/caac.21715

Bremnes, R. M., Al-Shibli, K., Donnem, T., Sirera, R., Al-Saad, S., Andersen, S., et al. (2011). The role of tumor-infiltrating immune cells and chronic inflammation at the tumor site on cancer development, progression, and prognosis: emphasis on non-small cell lung cancer. *J. Thorac. Oncol.* 6 (4), 824–833. doi:10.1097/jto. 0b013e3182037b76

Bruchard, M., Geindreau, M., Perrichet, A., Truntzer, C., Ballot, E., Boidot, R., et al. (2022). Recruitment and activation of type 3 innate lymphoid cells promote antitumor immune responses. *Nat. Immunol.* 23 (2), 262–274. doi:10.1038/s41590-021-01120-y

Chandrashekar, D. S., Bashel, B., Balasubramanya, S. A. H., Creighton, C. J., Ponce-Rodriguez, I., Chakravarthi, B. V. S. K., et al. (2017). UALCAN: A portal for facilitating tumor subgroup gene expression and survival analyses. *Neoplasia* 19 (8), 649–658. doi:10.1016/j.neo.2017.05.002

Chen, Y., Zhang, S., Wang, Q., and Zhang, X. (2017). Tumor-recruited M2 macrophages promote gastric and breast cancer metastasis via M2 macrophage-secreted CHI3L1 protein. *J. Hematol. Oncol.* 10 (1), 36. doi:10. 1186/s13045-017-0408-0

Chen, X., Li, A., Sun, B. F., Yang, Y., Han, Y. N., Yuan, X., et al. (2019). 5methylcytosine promotes pathogenesis of bladder cancer through stabilizing mRNAs. *Nat. Cell Biol.* 21 (8), 978–990. doi:10.1038/s41556-019-0361-y

Chowell, D., Yoo, S. K., Valero, C., Pastore, A., Krishna, C., Lee, M., et al. (2021). Improved prediction of immune checkpoint blockade efficacy across multiple cancer types. *Nat. Biotechnol.* 40, 499. doi:10.1038/s41587-021-01070-8

Clark, B. J. (2020). The START-domain proteins in intracellular lipid transport and beyond. *Mol. Cell. Endocrinol.* 504, 110704. doi:10.1016/j.mce.2020.110704

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/fcell.2022. 913684/full#supplementary-material

Commisso, C., Davidson, S. M., Soydaner-Azeloglu, R. G., Parker, S. J., Kamphorst, J. J., Hackett, S., et al. (2013). Macropinocytosis of protein is an amino acid supply route in Ras-transformed cells. *Nature* 497 (7451), 633–637. doi:10.1038/nature12138

Dai, S., Zeng, H., Liu, Z., Jin, K., Jiang, W., Wang, Z., et al. (2021). Intratumoral CXCL13+CD8+T cell infiltration determines poor clinical outcomes and immunoevasive contexture in patients with clear cell renal cell carcinoma. *J. Immunother. Cancer* 9 (2), e001823. doi:10.1136/jitc-2020-001823

Darvin, P., Toor, S. M., Sasidharan Nair, V., and Elkord, E. (2018). Immune checkpoint inhibitors: recent progress and potential biomarkers. *Exp. Mol. Med.* 50 (12), 1–11. doi:10.1038/s12276-018-0191-1

Daviet, L., Lehtonen, J. Y., Tamura, K., Griese, D. P., Horiuchi, M., and Dzau, V. J. (1999). Cloning and characterization of ATRAP, a novel protein that interacts with the angiotensin II type 1 receptor. *J. Biol. Chem.* 274 (24), 17058–17062. doi:10. 1074/jbc.274.24.17058

Downs-Canner, S. M., Meier, J., Vincent, B. G., and Serody, J. S. (2022). B cell function in the tumor microenvironment. *Annu. Rev. Immunol.* 40, 169. doi:10. 1146/annurev-immunol-101220-015603

Fan, H. N., Chen, Z. Y., Chen, X. Y., Chen, M., Yi, Y. C., Zhu, J. S., et al. (2022). METTL14-mediated m6A modification of circORC5 suppresses gastric cancer progression by regulating miR-30c-2-3p/AKT1S1 axis. *Mol. Cancer* 21 (1), 51. doi:10.1186/s12943-022-01521-z

Figueiredo, C., Camargo, M. C., Leite, M., Fuentes-Pananá, E. M., Rabkin, C. S., and Machado, J. C. (2017). Pathogenesis of gastric cancer: Genetics and molecular classification. *Curr. Top. Microbiol. Immunol.* 400, 277–304. doi:10.1007/978-3-319-50520-6_12

Fuji, K., Miwa, K., and Fujiwara, T. (2009). The intracellular transport of transporters: membrane trafficking of mineral transporters. *Curr. Opin. Plant Biol.* 12 (6), 699–704. doi:10.1016/j.pbi.2009.09.006

Fumet, J. D., Truntzer, C., Yarchoan, M., and Ghiringhelli, F. (2020). Tumour mutational burden as a biomarker for immunotherapy: Current data and emerging concepts. *Eur. J. Cancer* 131, 40–50. doi:10.1016/j.ejca.2020.02.038

Gan, X. X., Li, Y. Y., Li, S. J., Mo, S. S., Feng, J. H., Shen, F., et al. (2021). Significance of DMBT1 in papillary thyroid carcinoma concurrent with Hashimoto's thyroiditis. *Front. Oncol.* 11, 680873. doi:10.3389/fonc.2021.680873

Gonzalez, H., Hagerling, C., and Werb, Z. (2018). Roles of the immune system in cancer: from tumor initiation to metastatic progression. *Genes Dev.* 32 (19-20), 1267–1284. doi:10.1101/gad.314617.118

Green, B., Belcheva, A., Nepal, R. M., Boulianne, B., and Martin, A. (2011). The mismatch repair pathway functions normally at a non-AID target in germinal center B cells. *Blood* 118 (11), 3013–3018. doi:10.1182/blood-2011-03-345991

Hetz, C., and Saxena, S. (2017). ER stress and the unfolded protein response in neurodegeneration. *Nat. Rev. Neurol.* 13 (8), 477–491. doi:10.1038/nrneurol.2017.99

Hetz, C., Zhang, K., and Kaufman, R. J. (2020). Mechanisms, regulation and functions of the unfolded protein response. *Nat. Rev. Mol. Cell Biol.* 21 (8), 421–438. doi:10.1038/s41580-020-0250-z

Kanehisa, M., and Goto, S. (2000). KEGG: kyoto encyclopedia of genes and genomes. *Nucleic Acids Res.* 28 (1), 27-30. doi:10.1093/nar/28.1.27

Kanzler, H., Barrat, F. J., Hessel, E. M., and Coffman, R. L. (2007). Therapeutic targeting of innate immunity with Toll-like receptor agonists and antagonists. *Nat. Med.* 13 (5), 552–559. doi:10.1038/nm1589

Kent, W. J., Sugnet, C. W., Furey, T. S., Roskin, K. M., Pringle, T. H., Zahler, A. M., et al. (2002). The human genome browser at UCSC. *Genome Res.* 12 (6), 996–1006. doi:10.1101/gr.229102

Khan, N. H., Wang, D., Wang, W., Shahid, M., Khattak, S., Ngowi, E. E., et al. (2022). Pharmacological inhibition of endogenous hydrogen sulfide attenuates breast cancer progression. *Molecules* 27 (13), 4049. doi:10.3390/molecules27134049

Kim, H., and Chung, J. H. (2019). PD-L1 testing in non-small cell lung cancer: Past, present, and Future. *J. Pathol. Transl. Med.* 5354 (42), 199196–199206. doi:10. 4132/jptm.2019.04.24

Kim, J. Y., Banerjee, T., Vinckevicius, A., Luo, Q., Parker, J. B., Baker, M. R., et al. (2014). A role for WDR5 in integrating threonine 11 phosphorylation to lysine 4 methylation on histone H3 during androgen signaling and in prostate cancer. *Mol. Cell* 54 (4), 613–625. doi:10.1016/j.molcel.2014.03.043

Kim, S., Rhee, J. K., Yoo, H. J., Lee, H. J., Lee, E. J., Lee, J. W., et al. (2015). Bioinformatic and metabolomic analysis reveals miR-155 regulates thiamine level in breast cancer. *Cancer Lett.* 357 (2), 488–497. doi:10.1016/j.canlet.2014.11.058

Kim, T. W., Lee, S. Y., Kim, M., Cheon, C., and Ko, S. G. (2018). Kaempferol induces autophagic cell death via IRE1-JNK-CHOP pathway and inhibition of G9a in gastric cancer cells. *Cell Death Dis.* 9 (9), 875. doi:10.1038/s41419-018-0930-1

Kraehenbuehl, L., Weng, C. H., Eghbali, S., Wolchok, J. D., and Merghoub, T. (2022). Enhancing immunotherapy in cancer by targeting emerging immunomodulatory pathways. *Nat. Rev. Clin. Oncol.* 19 (1), 37–50. doi:10.1038/s41571-021-00552-7

Lam, H., McNeil, L. K., Starobinets, H., DeVault, V. L., Cohen, R. B., Twardowski, P., et al. (2021). An empirical antigen selection method identifies neoantigens that either elicit broad antitumor T-cell responses or drive tumor growth. *Cancer Discov.* 11 (3), 696–713. doi:10.1158/2159-8290.cd-20-0377

Lan, Q., Liu, P. Y., Haase, J., Bell, J. L., Hüttelmaier, S., and Liu, T. (2019). The critical role of RNA m6A methylation in cancer. *Cancer Res.* 79 (7), 1285–1292. doi:10.1158/0008-5472.can-18-2965

Latham, A., Srinivasan, P., Kemel, Y., Shia, J., Bandlamudi, C., Mandelker, D., et al. (2019). Microsatellite instability is associated with the presence of lynch syndrome pan-cancer. J. Clin. Oncol. 37 (4), 286–295. doi:10.1200/jco.18.00283

Li, T., Fan, J., Wang, B., Traugh, N., Chen, Q., Liu, J. S., et al. (2017). TIMER: A web server for comprehensive analysis of tumor-infiltrating immune cells. *Cancer Res.* 77 (21), e108–e110. doi:10.1158/0008-5472.can-17-0307

Li, G. M. (2008). Mechanisms and functions of DNA mismatch repair. *Cell Res.* 18 (1), 85–98. doi:10.1038/cr.2007.115

Lin, Y. H., Wu, M. H., Huang, Y. H., Yeh, C. T., Cheng, M. L., Chi, H. C., et al. (2018). Taurine up-regulated gene 1 functions as a master regulator to coordinate glycolysis and metastasis in hepatocellular carcinoma. *Hepatology* 67 (1), 188–203. doi:10.1002/hep.29462

Lin, T. C., Chien, W. C., Hu, J. M., Tzeng, N. S., Chung, C. H., Pu, T. W., et al. (2020). Risk of colorectal cancer in patients with alcoholism: a nationwide, population-based nested case-control study. *PLoS One* 15 (5), e0232740. doi:10. 1371/journal.pone.0232740

Liu, D., Keijzers, G., and Rasmussen, L. J. (2017). DNA mismatch repair and its many roles in eukaryotic cells. *Mutat. Research/Reviews Mutat. Res.* 773, 174–187. doi:10.1016/j.mrrev.2017.07.001

Liu, S., Zhao, W., Li, X., Zhang, L., Gao, Y., Peng, Q., et al. (2021). AGTRAP is a prognostic biomarker correlated with immune infiltration in hepatocellular carcinoma. *Front. Oncol.* 11, 713017. doi:10.3389/fonc.2021.713017

Locasale, J. W. (2013). Serine, glycine and one-carbon units: cancer metabolism in full circle. *Nat. Rev. Cancer* 13 (8), 572–583. doi:10.1038/nrc3557

Loftus, S. K., Baxter, L. L., Cronin, J. C., Fufa, T. D., and Pavan, W. J. (2017). Hypoxia-induced HIF1 attargets in melanocytes reveal a molecular profile associated with poor melanoma prognosis. *Pigment. Cell Melanoma Res.* 30 (3), 339–352. doi:10.1111/pcmr.12579

Loo, Y. M., and Gale, M. (2011). Immune signaling by RIG-I-like receptors. Immunity 34 (5), 680-692. doi:10.1016/j.immuni.2011.05.003

Lopez-Ilasaca, M., Liu, X., Tamura, K., and Dzau, V. J. (2003). The angiotensin II type I receptor-associated protein, ATRAP, is a transmembrane protein and a modulator of angiotensin II signaling. *MBoC* 14 (12), 5038–5050. doi:10.1091/mbc.e03-06-0383

Lunt, S. Y., and Vander Heiden, M. G. (2011). Aerobic glycolysis: meeting the metabolic requirements of cell proliferation. *Annu. Rev. Cell Dev. Biol.* 27, 441–464. doi:10.1146/annurev-cellbio-092910-154237

Maeda, A., Tamura, K., Wakui, H., Dejima, T., Ohsawa, M., Azushima, K., et al. (2013). Angiotensin receptor-binding protein ATRAP/Agtrap inhibits metabolic dysfunction with visceral obesity. *J. Am. Heart Assoc.* 2 (4), e000312. doi:10.1161/JAHA.113.000312

Malmberg, K. J., Carlsten, M., Björklund, A., Sohlberg, E., Bryceson, Y. T., and Ljunggren, H. G. (2017). Natural killer cell-mediated immunosurveillance of human cancer. *Seminars Immunol.* 31, 20–29. doi:10.1016/j.smim.2017.08.002

Mandal, R., Samstein, R. M., Lee, K. W., Havel, J. J., Wang, H., Krishna, C., et al. (2019). Genetic diversity of tumors with mismatch repair deficiency influences anti-PD-1 immunotherapy response. *Science* 364 (6439), 485–491. doi:10.1126/science. aau0447

Mao, X., Xu, J., Wang, W., Liang, C., Hua, J., Liu, J., et al. (2021). Crosstalk between cancer-associated fibroblasts and immune cells in the tumor microenvironment: new findings and future perspectives. *Mol. Cancer* 20 (1), 131. doi:10.1186/s12943-021-01428-1

Mas-Ponte, D., and Supek, F. (2020). DNA mismatch repair promotes APOBEC3-mediated diffuse hypermutation in human cancers. *Nat. Genet.* 52 (9), 958–968. doi:10.1038/s41588-020-0674-6

Mayakonda, A., Lin, D. C., Assenov, Y., Plass, C., and Koeffler, H. P. (2018). Maftools: efficient and comprehensive analysis of somatic variants in cancer. *Genome Res.* 28 (11), 1747–1756. doi:10.1101/gr.239244.118

Michael-Robinson, J. M., Biemer-Hüttmann, A., Purdie, D. M., Walsh, M. D., Simms, L. A., Biden, K. G., et al. (2001). Tumour infiltrating lymphocytes and apoptosis are independent features in colorectal cancer stratified according to microsatellite instability status. *Gut* 48 (3), 360–366. doi:10.1136/gut.48.3.360

Miles, L. A., Bowman, R. L., Merlinsky, T. R., Csete, I. S., Ooi, A. T., and Durruthy-Durruthy, R. (2021). PCSK9 causes MHC class I lysosomal degradation to promote tumor growth. *Cancer Discov.* 11 (1), 13. doi:10.1158/2159-8290.CD-RW2020-167

Myers, L., and Sirois, M. J. (2004). "Spearman correlation coefficients, differences between," in *Encyclopedia of statistical sciences*. doi:10.1002/0471667196.ess5050

Nagarsheth, N., Wicha, M. S., and Zou, W. (2017). Chemokines in the cancer microenvironment and their relevance in cancer immunotherapy. *Nat. Rev. Immunol.* 17 (9), 559–572. doi:10.1038/nri.2017.49

Nishimura, T., and Stefan, C. J. (2020). Specialized ER membrane domains for lipid metabolism and transport. *Biochim. Biophys. Acta Mol. Cell Biol. Lipids* 1865 (1), 158492. doi:10.1016/j.bbalip.2019.07.001

Ohsawa, M., Tamura, K., Wakui, H., Maeda, A., Dejima, T., Kanaoka, T., et al. (2014). Deletion of the angiotensin II type 1 receptor-associated protein enhances renal sodium reabsorption and exacerbates angiotensin II-mediated hypertension. *Kidney* 1nt. 86 (3), 570–581. doi:10.1038/ki.2014.95

Pan, Y., Yu, Y., Wang, X., and Zhang, T. (2020). Tumor-associated macrophages in tumor immunity. *Front. Immunol.* 11, 583084. doi:10.3389/fimmu.2020.583084

Patel, S., Kim, J., Herrera, M., Mukherjee, A., Kabanov, A. V., and Sahay, G. (2019). Brief update on endocytosis of nanomedicines. *Adv. Drug Deliv. Rev.* 144, 90–111. doi:10.1016/j.addr.2019.08.004

Pestana, R. C., Beal, J. R., Parkes, A., Hamerschlak, N., and Subbiah, V. (2022). Impact of tissue-agnostic approvals for patients with sarcoma. *Trends Cancer* 8 (2), 135–144. doi:10.1016/j.trecan.2021.11.007

Pugh-Toole, M., Nicolela, A. P., Nersesian, S., Leung, B. M., and Boudreau, J. E. (2022). Natural killer cells: the missing link in effective treatment for high-grade serous ovarian carcinoma. *Curr. Treat. Options Oncol.* 23, 210. doi:10.1007/s11864-021-00929-x

Reck, M., Rodríguez-Abreu, D., Robinson, A. G., Hui, R., Csőszi, T., Fülöp, A., et al. (2019). Updated analysis of KEYNOTE-024: Pembrolizumab versus platinumbased chemotherapy for advanced non-small-cell lung cancer with PD-L1 tumor proportion score of 50% or greater. *J. Clin. Oncol.* 37 (7), 537–546. doi:10.1200/JCO. 18.00149

Revelle, W., and Revelle, M. W. (2015). Package 'psych'. The comprehensive R archive network, 337–338.

Sahin, I. H., Akce, M., Alese, O., Shaib, W., Lesinski, G. B., El-Rayes, B., et al. (2019). Immune checkpoint inhibitors for the treatment of MSI-H/MMR-D colorectal cancer and a perspective on resistance mechanisms. *Br. J. Cancer* 121 (10), 809–818. doi:10.1038/s41416-019-0599-y

Sanz-Pamplona, R., Gil-Hoyos, R., López-Doriga, A., Alonso, M. H., Aussó, S., Molleví, D. G., et al. (2016). Mutanome and expression of immune response genes in microsatellite stable colon cancer. *Oncotarget* 7 (14), 17711–17725. doi:10.18632/ oncotarget.7293

Scharping, N. E., Rivadeneira, D. B., Menk, A. V., Vignali, P. D. A., Ford, B. R., Rittenhouse, N. L., et al. (2021). Mitochondrial stress induced by continuous stimulation under hypoxia rapidly drives T cell exhaustion. *Nat. Immunol.* 22 (2), 205–215. doi:10.1038/s41590-020-00834-9

Schmid, P., Adams, S., Rugo, H. S., Schneeweiss, A., Barrios, C. H., Iwata, H., et al. (2018). Atezolizumab and nab-paclitaxel in advanced triple-negative breast cancer. *N. Engl. J. Med.* 379 (22), 2108–2121. doi:10.1056/nejmoa1809615

Schumacher, T. N., and Schreiber, R. D. (2015). Neoantigens in cancer immunotherapy. Science 348 (6230), 69-74. doi:10.1126/science.aaa4971

Song, B. N., Kim, S. K., and Chu, I. S. (2017). Bioinformatic identification of prognostic signature defined by copy number alteration and expression of CCNE1 in non-muscle invasive bladder cancer. *Exp. Mol. Med.* 49 (1), e282. doi:10.1038/emm.2016.120

Subramanian, A., Tamayo, P., Mootha, V. K., Mukherjee, S., Ebert, B. L., Gillette, M. A., et al. (2005). Gene set enrichment analysis: a knowledge-based approach for interpreting genome-wide expression profiles. *Proc. Natl. Acad. Sci. U.S.A.* 102 (43), 15545–15550. doi:10.1073/pnas.0506580102

Sun, T., Yang, W., Toprani, S. M., Guo, W., He, L., DeLeo, A. B., et al. (2020). Induction of immunogenic cell death in radiation-resistant breast cancer stem cells by repurposing anti-alcoholism drug disulfiram. *Cell Commun. Signal* 18 (1), 36. doi:10.1186/s12964-019-0507-3

Szklarczyk, D., Gable, A. L., Lyon, D., Junge, A., Wyder, S., Huerta-Cepas, J., et al. (2019). STRING v11: protein-protein association networks with increased coverage, supporting functional discovery in genome-wide experimental datasets. *Nucleic Acids Res.* 47 (D1), D607–D613. doi:10.1093/nar/gky1131

Tamura, K., Wakui, H., Maeda, A., Dejima, T., Ohsawa, M., Azushima, K., et al. (2013). The physiology and pathophysiology of a novel angiotensin receptorbinding protein ATRAP/Agtrap. *Cpd* 19 (17), 3043–3048. doi:10.2174/ 1381612811319170010

Tamura, K., Azushima, K., Kinguchi, S., Wakui, H., and Yamaji, T. (2022). ATRAP, a receptor-interacting modulator of kidney physiology, as a novel player in blood pressure and beyond. *Hypertens. Res.* 45 (1), 32–39. doi:10.1038/s41440-021-00776-1

Tan, Y., Dourdin, N., Wu, C., De Veyra, T., Elce, J. S., and Greer, P. A. (2006). Ubiquitous calpains promote caspase-12 and JNK activation during endoplasmic reticulum stress-induced apoptosis. *J. Biol. Chem.* 281 (23), 16016–16024. doi:10. 1074/jbc.m601299200

Tang, Z., Kang, B., Li, C., Chen, T., and Zhang, Z. (2019). GEPIA2: an enhanced web server for large-scale expression profiling and interactive analysis. *Nucleic Acids Res.* 47 (W1), W556–W560. doi:10.1093/nar/gkz430

The Gene Ontology Consortium (2019). The gene Ontology resource: 20 years and still GOing strong. *Nucleic Acids Res.* 47 (D1), D330–D338. doi:10.1093/nar/gky1055

Thul, P. J., and Lindskog, C. (2018). The human protein atlas: a spatial map of the human proteome. *Protein Sci.* 27 (1), 233–244. doi:10.1002/pro.3307

Veneziani, I., Alicata, C., Pelosi, A., Landolina, N., Ricci, B., D'Oria, V., et al. (2022). Toll-like receptor 8 agonists improve NK-cell function primarily targeting CD56brightCD16– subset. *J. Immunother. Cancer* 10 (1), e003385. doi:10.1136/jitc-2021-003385

Wang, Q., Wang, H., Jia, Y., Pan, H., and Ding, H. (2017). Luteolin induces apoptosis by ROS/ER stress and mitochondrial dysfunction in gliomablastoma. *Cancer Chemother. Pharmacol.* 79 (5), 1031–1041. doi:10.1007/s00280-017-3299-4

Wang, X., Qiu, Y., Wang, M., Zhang, C., Zhang, T., Zhou, H., et al. (2020). Endocytosis and organelle targeting of nanomedicines in cancer therapy. *Int. J. Nanomedicine* 15, 9447–9467. doi:10.2147/ijn.s274289

Wang, Y., Wang, J., Li, X., Xiong, X., Wang, J., Zhou, Z., et al. (2021). N1methyladenosine methylation in tRNA drives liver tumourigenesis by regulating cholesterol metabolism. *Nat. Commun.* 12 (1), 6314. doi:10.1038/s41467-021-26718-6 Wang, Z., Li, D., Mo, L., Liang, S., Liao, X., Guo, S., et al. (2022). Low-dose cadmium exposure promotes osteoclastogenesis by enhancing autophagy via inhibiting the mTOR/p70S6K1 signaling pathway. *Toxicol. Lett.* 367 (22), 900150–900183. doi:10.1016/j.toxlet.2022.07.005

Xiao, F., Li, J., Huang, K., Li, X., Xiong, Y., Wu, M., et al. (2021). Macropinocytosis: mechanism and targeted therapy in cancers. *Am. J. Cancer Res.* 11 (1), 14–30.

Xiong, Z., Xiong, Y., Liu, H., Li, C., and Li, X. (2020). Identification of purity and prognosis related gene signature by network analysis and survival analysis in brain lower grade glioma. *J. Cell. Mol. Med.* 24 (19), 11607–11612. doi:10.1111/jcmm. 15805

Xu, L., Chen, X., Jiang, H., Xu, J., Wang, L., and Sun, Y. (2021). NDUFC1 is upregulated in gastric cancer and regulates cell proliferation, apoptosis, cycle and migration. *Front. Oncol.* 11, 709044. doi:10.3389/fonc.2021.709044

Yarchoan, M., Johnson, B. A., Lutz, E. R., Laheru, D. A., and Jaffee, E. M. (2017). Targeting neoantigens to augment antitumour immunity. *Nat. Rev. Cancer* 17 (4), 209–222. doi:10.1038/nrc.2016.154

Yin, M., Li, X., Tan, S., Zhou, H. J., Ji, W., Bellone, S., et al. (2016). Tumorassociated macrophages drive spheroid formation during early transcoelomic metastasis of ovarian cancer. *J. Clin. Invest.* 126 (11), 4157–4173. doi:10.1172/ jci87252

You, K., Wang, L., Chou, C. H., Liu, K., Nakata, T., Jaiswal, A., et al. (2021). QRICH1 dictates the outcome of ER stress through transcriptional control of proteostasis. *Science* 371 (6524), eabb6896. doi:10.1126/science.abb6896

Yu, G., Wang, L. G., Han, Y., and He, Q. Y. (2012). clusterProfiler: an R package for comparing biological themes among gene clusters. *OMICS A J. Integr. Biol.* 16 (5), 284–287. doi:10.1089/omi.2011.0118

Zeng, R., Lyu, Y., Zhang, G., Shou, T., Wang, K., Niu, H., et al. (2018). Positive effect of ROR γ t on the prognosis of thyroid papillary carcinoma patients combined with Hashimoto's thyroiditis. *Am. J. Transl. Res.* 10 (10), 3011–3024.

Zeng, H., Li, H., Zhao, Y., Chen, L., and Ma, X. (2019). Transcripto-based network analysis reveals a model of gene activation in tongue squamous cell carcinomas. *Head Neck* 41 (12), 4098–4110. doi:10.1002/hed.25952

Zeng, D., Ye, Z., Shen, R., Yu, G., Wu, J., Xiong, Y., et al. (2021). IOBR: Multi-Omics immuno-oncology biological research to decode tumor microenvironment and signatures. *Front. Immunol.* 12, 687975. doi:10.3389/ fimmu.2021.687975

Zhang, X., Hu, P., Ding, S. Y., Sun, T., Liu, L., Han, S., et al. (2019). Induction of autophagy-dependent apoptosis in cancer cells through activation of ER stress: an uncovered anti-cancer mechanism by anti-alcoholism drug disulfiram. *Am. J. Cancer Res.* 9 (6), 1266–1281.

Zhu, C., Mustafa, D., Zheng, P. P., van der Weiden, M., Sacchetti, A., Brandt, M., et al. (2017). Activation of CECR1 in M2-like TAMs promotes paracrine stimulation-mediated glial tumor progression. *Neuro Oncol.* 19 (5), 648–659. doi:10.1093/neuonc/now251

Zhu, L., Liu, S., Liao, Y. F., Sheng, Y. M., He, J. C., Cai, Z. X., et al. (2022). Calycosin suppresses colorectal cancer progression by targeting $ER\beta$, upregulating PTEN, and inhibiting PI3K/Akt signal pathway. *Cell Biol. Int.* 46 (9), 1367–1377. doi:10.1002/cbin.11840

Zlotnik, A., Burkhardt, A. M., and Homey, B. (2011). Homeostatic chemokine receptors and organ-specific metastasis. *Nat. Rev. Immunol.* 11 (9), 597–606. doi:10. 1038/nri3049

Glossary

AGTRAP angiotensin II receptor-associated protein ALL acute lymphoblastic leukemia AT1R Ang II type 1 receptor BLCA bladder urothelial carcinoma BRCA breast invasive carcinoma CESC cervical squamous cell carcinoma and endo-cervical adenocarcinoma CHOL cholangiocarcinoma CIs confidence intervals COAD colon adenocarcinoma CPTAC Clinical Proteomic Tumor Analysis Consortium DFS disease-free survival DSS disease-specific survival EPCAM epithelial cell adhesion molecule ER endoplasmic reticulum ESCA esophageal carcinoma FISH fluorescence in situ hybridization GBM glioblastoma multiforme, GEPIA2 Gene Expression Profiling Interactive Analysis, version 2 GO Gene Ontology GSEA Gene set enrichment analysis GTEx Genotype-Tissue Expression HNSC head and neck squamous cell carcinoma HRs hazard ratios ICIs immune checkpoint inhibitors KEGG Kyoto Encyclopedia of Genes and Genomes KICH kidney chromophobe **KIPAN** Pan-kidney KIRC kidney renal clear cell carcinoma KIRP kidney renal papillary cell carcinoma LAML acute myeloid leukemia LGG brain lower grade glioma LIHC liver hepatocellular carcinoma LUAD lung adenocarcinoma LUSC lung squamous cell carcinoma

m(1)A N1-methyladenosine m(5)C 5-methylcytosine **m(6)A** N6-methyladenosine MAPK mitogen-activated protein kinase MESO mesothelioma MHC-1 Class I major histocompatibility complex MLH1 MutL protein homolog 1 MMR mismatch repair MSH2 MutS protein homolog 2 MSH6 MutS homolog 6 MSI-H high microsatellite instability NB neuroblastoma NF-KB factor-kappa B subunit 1 NK natural killer **NMs** nanomedicines OS overall survival OV ovarian serous cystadenocarcinoma PAAD pancreatic adenocarcinoma PCPG pheochromocytoma and paraganglioma PFS progression-free survival PMS2 PMS1 homolog 2 PPI protein-protein interaction PRAD prostate adenocarcinoma RAS renin-angiotensin system READ rectum adenocarcinoma SARC sarcoma SKCM skin cutaneous melanoma STAD stomach adenocarcinoma STES stomach and esophageal carcinoma TAMs tumor-associated macrophages TARGET Therapeutically Applicable Research to Generate Effective Treatments TCGA The Cancer Genome Atlas TCGA-ACC The Cancer Genome Atlas Adrenocortical Carcinoma TCGA-USC The Cancer Genome Atlas Uterine Serous Carcinoma TGCT testicular germ cell tumors

- THCA thyroid carcinoma THPA The Human Protein Atlas TILs tumor-infiltrating lymphocytes TIMER Tumor Immune Estimation Resource TMB tumor burden TME tumor microenvironment TPM transcripts per million
- UCEC uterine corpus endometrial carcinoma UCS uterine carcinosarcoma UCSC University of California, Santa Cruz UPR unfolded protein response UVM uveal melanoma WT Wilms' tumor