CENPF/CDK1 Signaling Pathway Enhances the Progression of Adrenocortical Carcinoma by Regulating the G2/M-phase Cell Cycle

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Research Article

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CENPF/CDK1 Signaling Pathway enhances the Progression of Adrenocortical Carcinoma by Regulating the G2/M-phase Cell Cycle

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Abstract

Adrenocortical carcinoma (ACC) is an aggressive and rare malignant tumor and prone to local invasion and metastasis. While, overexpressed Centromere Protein F (CENPF) is closely related to oncogenesis of various neoplasms, including ACC. However, the prognosis and exact biological function of CENPF in ACC remains largely unclear. In present essay, the expression of CENPF in human ACC samples, GEO and TCGA databases depicted that CENPF were overtly hyper-expressed in ACC patients and positively correlated with tumor stage. The aberrant expression of CENPF was significantly correlated with unfavorable overall survival (OS) in ACC patients. Then, the application of gene-set enrichment analysis (GSEA) declared that CENPF was mainly involved in the G2/M-phase mediated cell cycle and p53 signaling pathway. Further, a small RNA interference experiment was conducted to demonstrate that the interaction between CENPF and CDK1 enhanced the G2/M-phase transition of mitosis, cell proliferation and might induce p53 mediated anti-tumor effect in human ACC cell line, SW13 cells. Lastly, two available therapeutic strategies, including immunotherapy and chemotherapy, have been further probed. Immune infiltration analysis highlighted that ACC patients with high CENPF expression harbored significantly different immune cell populations, and high TMB/MSI score. Then, the gene-drug interaction network stated that CENPF inhibitors, such as Cisplatin, Sunitinib, and Etoposide, might serve as potential drugs for the therapy of ACC. Briefly, CENPF and related genes might be served as a novel prognostic biomarker or latent therapeutic target for ACC patients.

Key Words: adrenocortical carcinoma, ACC, CENPF, CDK1, p53, cell cycle

Background

Adrenocortical carcinoma (ACC), an invasive and rare malignant neoplasm originated from the adrenal cortex, is prone to local invasion and metastasis through blood and lymph nodes, accounting for 14% of primary adrenal incidentalomas[1]. The overall incidence of ACC reaches peak in childhood and adults (40-60 years old) with a bimodal age distribution. In the United States, general cancer incidence is about 4500 cases/million/year in adults and 150 cases/million/year in children[2, 3]. The median overall survival is about 3.5 years, and the 5-year survival rate is about 35% for ACC with locally advanced stages, and 0–28% for ACC with metastases[4, 5]. It indicates that ACC is characterized with tumor heterogeneity and harbors a dismal prognosis. At present, effective drugs for the therapies of ACC are rare. From 1914 to now, surgery and mitotane/platinum-based chemotherapy are still the only available or effective therapeutic strategies[1]. ACC is still a homogeneous class with unique biological characteristics from a pan-cancer landscape, even if ACC emerges heterogeneous molecular profiles. Hitherto, several biomarkers associated with diagnosis, prognosis or therapy, which might exert vital roles in determining the aggressiveness of ACC, have been reported. Molecular studies have demonstrated that TP53 [6, 7] and CTNNB1[8] mutations serve as the most common driver genes for ACC patients. About 10% ACC patients from the TCGA datasets had a pathogenetic TP53 variant[8]. Sbiera S et al. proposed that SOAT1 may serve as
molecular target of mitotane in ACC[9]. Accumulative literatures have demonstrated that inhibition of WNT–β-catenin signaling pathway[10], cell- cycle progression through down-regulation of p53–RB[11], defects in mismatch repair related enzymes (dMMR) [12], DNA methylation[13], and abnormal maintenance of telomere (ATRX[14, 15], DAX[16] and TERT[8, 17]), were closely bound up with the deterioration and poor prognosis of ACC. In addition, the unique feature of ACC is steroid differentiation, which has been viewed as a potential therapeutic target. Some drugs are undergoing preclinical or clinical research[18]. Despite the above biomarkers of ACC have attracted soaring attention, the relevant research is still in the stage of experimental research or clinical trial. Hence, it is necessary to scout masked bio-targets to forecast the prognosis in early stage or adopt new treatment strategies.

The centromeric protein F (CENPF), a mitotic centromere protein, is involved in chromosome segregation during mitosis of cell cycle and expressed in a cell cycle- related pattern. The expression peak of CENPF is in the cell cycle of G2/M phase. The aberrant expression of CENPF has been found in multiple malignancies, including hepatocellular carcinoma (HCC) [19-21], prostate cancer[22] and other carcinomas [23, 24], but few literatures about ACC. Specifically, the latest studies have acclaimed that overexpression of CENPF markedly promotes the cell proliferation in HCC[20] and induce tumor metastasis in prostate cancer with a poor prognosis[25]. Lin S et al. have reported that the signal transduction mediated by CENPF promoted the recurrence and metastasis of prostate cancer[26]. Consequently, CENPF might be a meaningful biomarker for inferring the prognosis of various cancers, including ACC; Moreover, screening the potential molecular mechanisms of CENPF mediated carcinogenic or anti-tumor signaling pathways may furnish new therapeutic strategies for ACC. Glancing at the expression profile of CENPF in various cancers, we found that CENPF was overexpressed in ACC and negatively correlated with prognosis. However, the expression level, genetic changes, biological process, molecular mechanism and prognosis of CENPF in ACC have not been further elucidated.

As one of powerful weapons for biomedical research, the gene-chip and high-throughput sequencing technology has greatly enriched and promoted the research of cancer genomics[27-29]. In this essay, combined with the IHC detection of human samples, cell line experiment and the analysis of GEO and TCGA databases, it was found that CENPF was evidently overexpressed, associated with dismal prognosis, involved in cell cycle, and related signaling pathways in ACC. The results of this study will help to broaden the understanding of ACC, comprehensively analyze the correlation between CENPF and the pathogenesis and progress of ACC, and furnish beneficial clues for the initiation and development of ACC.

Results

**CENPF is overexpressed in ACC, but not in adrenocortical adenoma or normal adrenal cortex tissues**

Significant differences were found in morphology and histology among normal adrenal cortex, adrenocortical adenoma (benign tumor) and ACC (malignant tumor). Firstly, gross specimen photography showed that for adrenocortical adenoma, the nodule section was golden and hard; For ACC, the capsule was intact or defective, some areas were gray red and necrotic (Figure 1A). Results of IHC-P showed that CENPF was mostly upregulated in ACC compared with adrenocortical adenoma samples or normal adrenal cortex tissues, but no significant difference was discovered between adrenocortical adenoma and normal tissues (P <0.001, Figure 1B, C). Furthermore, the results were validated in public database, including GEO and TCGA datasets, for ACC patients. The
mRNA expression of CENPF was dramatically overexpressed in GSE90713 (P < 0.01, Figure 1D). But interestingly, there was no significant difference in the expression of this gene in GSE19750 (P=0.11, Figure 1E). This might be in case of too small sample size in normal adrenocortical tissues (n=4). Meanwhile, the controversy attracted us to further explore. Moreover, correlation analysis by the GEPIA database showed that the mRNA expression of CENPF was evidently overexpressed in ACC tumors compared with that in normal ones (P < 0.001, Figure 1F), and was positively correlated with the tumor stage for ACC (P < 0.001, Figure 1G).

Additionally, Ki67, as a cell proliferation index, was investigated in ACC. According to the proportion of Ki67 positive cells in ACC, the relationship between the expression of CENPF and cell proliferation was further discussed. As results, the expression of CENPF, namely the proportion of CENPF positive cells, was positively correlated with the proportion of Ki67 positive cells (P =0.008, R=0.85, Figure 2A-B). Moreover, the expression of CENPF was positively correlated with that of MKI67 expression based on TCGA-ACC dataset (P =0, R=0.91, Figure2C). It indicated that the expression of CENPF was closely related to the cell proliferation in patients with ACC.

Analysis of CENPF in clinicopathological parameters for ACC patients

For figuring out the correlation between CENPF expression and clinicopathological parameters in ACC, all ACC patients were divided into two subgroups, including 39 CENPF<sub>low</sub> and 40 CENPF<sub>high</sub> group according to the median cutoffs (Figure 3A). As presented in Table 1, the expression of CENPF was not correlated with age, gender, laterality, and mitotane therapy (P>0.05). Compared with CENPF<sub>low</sub>, ACC patients with CENPF<sub>high</sub> suffered manifestly advanced pathological stages (especially stage III and IV, P < 0.001). In tumor status, new events or residual tumor, more ACC cases with CENPF<sub>high</sub> were prone to lose the chance of surgery, relapse or form residual tumor than CENPF<sub>low</sub> (P < 0.001). Additionally, CENPF<sub>high</sub> group showed a higher Weiss score (especially Weiss score 6–9, P = 0.002). Moreover, the sanguini diagram described the distribution of CENPF in age, gender, pTNM stage, and status (Figure 3B). Further, the correlation between CENPF and relevant clinical parameters, including age, gender, pTNM stage, etc. on the OS of ACC patients were identified via the uni-cox and multi-cox regression analysis. As results, the CENPF expression and pTNM stage were closely related to the OS of ACC patients in uni-cox analysis (Figure 3C, all P<0.05), and the CENPF expression and pTNM stage could be served as independent prognostic factors for ACC patients in multi-cox analysis (Figure 3D, all P<0.05). Finally, the survival rate of 1-year, 2- year or 3- year in one ACC patients related to high CENPF expression were evaluated by Nomogram (Figure 3E).

Overexpressed CENPF is associated with dismal survival in ACC patients

According to the survival rate analysis, upregulation of CENPF expression predicts poor OS (HR = 8.66, log rank P =1.8e-05; Figures 4A) and PFS (HR = 4.11, log rank P = 6.68e-05; Figures 4C) in ACC. ROC curve analyses demonstrated that AUC of CENPF at 1-year, 3-year, and 5-year for OS (Figure 3B) and PFS (Figure 3D) pointed out that the expression of CENPF has a good predictive effect on the prognosis of ACC (all AUC value >0.75). Additionally, subgroup analyses indicated that overexpression of CENPF in ACC patients acted as a risk factor for 30-month, 60-month and 120-month OS (all log rank P ≤0.001, Figures 4E-G).

GSEA analyses reveal masked molecular mechanisms of CENPF in tumorigenesis and progression
To uncover the potential roles of *CENPF* in cancer-related signaling pathways, GSEA was carried out to explain the gene expression profiles of ACC samples with *CENPF*\(^{\text{low}}\) and *CENPF*\(^{\text{high}}\). According to GSEA analysis of KEGG pathway, *CENPF*\(^{\text{high}}\) patients were mainly enriched in cell cycle, DNA replication, nucleotide excision repair, and p53 signaling pathway, etc. (Figure 5A, Supplementary Table 5). Then, GSEA enrichment analyses of Biocarta pathway and Hallmark description suggested that cell cycle (NES=2.03, \(P=0\); especially G2-phase of mitosis, NES=1.96, \(P=0\)), G2/M checkpoint (NES=2.01, \(P=0\)), and E2F targets (NES=2.04, \(P=0\)) were mostly enriched in ACC patients with *CENPF*\(^{\text{high}}\), which implied that *CENPF* might regulate cell cycle by interacting with E2Fs proteins (Figure 5B, Supplementary Tables 6-8). Additionally, relevant studies have proclaimed that E2F1 played the role as a transcription factor of *CENPF* in NCI-60 cell line [30] and regulated cell cycle of the G2/M-phase transition [31, 32]. Thus, we supposed that *CENPF* may regulate cell cycle by interacting with E2F1 in ACC. Venn graph suggested that 20 genes (CDK1, CDC25A, CDKN1A, E2F1, CCNB1, CDKN2D, ATR, TP53, ATM, CDK2, CCNE1, CDKN2A, TFDP1, CCND1, CDK6, CCNA1, CDK4, CDKN1B, CDKN2B and RB1) were co-expressed in at least two groups (Figure 5C). It has been reported that *CENPF* mainly regulated the spindle separation in mitosis[33]. As listed in Figure 5D and Supplementary Table 7, “spindle midzone” was the top 1 term in GESA analysis of GO terms. By screening the correlation, expression and prognosis of these 20 genes, we picked out that CDK1, CCNB1 and E2F1 were mostly overexpressed in ACC samples compared with normal ones (Figure 5F, all \(P<0.05\)), and positively relevant with the expression of *CENPF* (Figure 5E, all \(P<0.01\)). The overexpressed CDK1, CCNB1 and E2F1 were also related to negative OS of ACC patients (Figure 5G, all \(P<0.01\)). In addition, PPI network revealed that close interactions were found among CENPF, CDK1 and CCNB1 (Figure 5H). Thus, it suggested that CENPF regulated cell cycle by interacting with CDK1, E2F1 and CCNB1 in ACC. The overexpressed *CENPF* might play pivotal roles in the regulation of G2/M-phase transition mediated cell cycle in the progression of ACC.

**CENPF interference affects the regulation of cell cycle in SW13 cells**

In order to further search the role of *CENPF* in ACC, *CENPF* siRNAs (siCENPF) were conducted via ACC cell line, human SW13 cells, as *in vitro* experiments. The expression of CENPF was overtly inhibited in siCENPF, compared with siNC (Figures 6A-B). In cell proliferation assay, *CENPF* interference evidently inhibited cell proliferation (Figure 6C) and cell migration (Figure 6D) in human SW13 cells after cells were transfected for 48 h. Inhibition of *CENPF* distinctly reduced adhesion between tumor cells and matrix (Figure 6E) and cell invasion (Figure 6F) in human SW13 cells. In addition, flow cytometry assay proclaimed that down-regulation of *CENPF* overtly intensified cell apoptosis, including early and late cell apoptosis (18.46% vs 36.56%, Figure 6G). Then, *CENPF* interference induced cell cycle arrest at the G2/M-phase transition in human SW13 cells (28.2% vs 6.4%, Figure 6H). As analyzed above in Figure 5, the expression of *CENPF* was closely related to *CDK1* in ACC. Therefore, SiCDK1 was designed to probe the correlation between *CENPF* and *CDK1* (Supplementary Figures 1A-B). Down-regulation of *CDK1* significantly suppressed cell proliferation (Supplementary Figure 1C) and cell mobility (Figure 6D) in human SW13 cells after cells were transfected 48 h. Down-regulation of *CDK1* significantly restrained adhesion between tumor cells and matrix (Figure 6E) and cell invasion (Figure 6F) in human SW13 cells. Additionally, the flow cytometry experiment depicted that down-regulation of *CDK1* overtly augmented cell apoptosis, including early and late cell apoptosis (18.46% vs 28.5%, Figure 6G). Lastly, *CDK1* interference blocked cell cycle at the G2/M-phase transition in human SW13 cells (28.2% vs 16.2%, Figure 6H).

**CENPF regulates CDK1 expression and p53 signaling in human SW13 cells**

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The molecular mechanism of CENPF was further explored in SW13 cells. As represented in Figure 6, inhibition of the expression of CENPF resulted in the decrease of CDK1 expression (Figure 6I), while inhibition of CDK1 had no significant effect on the expression of CENPF (Figure 6J). It suggested that CENPF might regulate the expression of CDK1. Furthermore, the GSEA enrichment analysis disclosed that p53 signaling pathway might serve as masked molecular mechanism of CENPF involved in the occurrence and development of human ACC (Figure 5A). Therefore, the phosphorylation of p53 (P-p53), the expression levels of p53, p21 and Bax were detected by Western blotting in SW13 cells. As the interference of CENPF and CDK1 by siRNA, the expression levels of P-p53, p21 and Bax were significantly decreased compared with the siNC in SW13 cells (Figure 6I-J). Therefore, inhibition of CENPF downregulated the expression of CDK1, suppressed the phosphorylation of p53 and altered downstream protein levels, including apoptosis and cell cycle related proteins.

The potential therapeutic strategies for the therapy of ACC

Next, the available chemical drugs and treatment strategies (including immunotherapy) that may be favorable for the treatment of ACC were further explored. Firstly, the immune cell infiltration in ACC with CENPF high and CENPF low were analyzed to investigate immunotherapy. All the 79 ACC tissues from TCGA database were divided to two groups, including 40 CENPF high and 39 CENPF low tissues by median value. CIBERSORTx was applied to analyze immune cell infiltration, including 22 cell types with low and high expression of CENPF. P<0.05 was statistically significant. The normalized by 'Limma' packages and then filtered groups including 7 CENPF low and 6 CENPF high ACC tissues were identified for further analysis. The expression of 22 immune cell populations for each sample were presented as heatmaps (Figure 7A). The immune enrichment score was dramatically various between 6 CENPF high and 7 CENPF low samples. As shown in Figure 7B, follicular helper T cells, M0 macrophages, Eosinophils was mainly enriched in ACC with CENPF high, compared with CENPF low group. However, the immune enrichment score of gamma delta T cells, monocytes, and M2 macrophages were much lower in CENPF high, compared with CENPF low group. The principal component analysis (PCA) showed that manifest difference was probed in immune infiltration between CENPF high and CENPF low ACC samples (Figure 7C). It might indicate that the immune heterogeneity of the CENPF was significant in ACC patients. Otherwise, the mRNA expression of LAG3, CTLA4, PD-1, PD-L1 and HAVCR2, tightly related to tumor immunosuppression or tolerance, were not significant difference in ACC with CENPF high, compared with CENPF low (Figure 7D, all P>0.05). Moreover, the correlation between the expression of CENPF and MSI or TMB were analyzed to unraveled that CENPF expression was significantly associated with MSI score (Figure 7E, P=0.034) and TMB score (Figure 7F, P<0.01) in ACC. It showed that overexpression of CENPF might augment the accumulation of abnormal gene mutation, namely TMB, in the process of DNA replication in ACC. Then, the gene-drug interaction network indicated that a variety of drugs could affect the expression levels of CENPF in mRNA or protein level (Figure 7G). For example, Cisplatin, Sunitinib, and Etoposide could restrain CENPF expression level while Paclitaxel and Genistein could induce CENPF expression level. Generally, all these CENPF inhibitors were deemed as potential targets for the treatment of ACC.

Discussion

Accumulative literatures have asserted that CENPF is closely related to in cell proliferation as well as tumorigenesis[19, 20, 33]. In recent years, evidences have arisen that the overexpression of CENPF was a frequent behavior in different malignant tumors and was closely related to the tumor deterioration and dismal prognosis in multiple neoplasms, including HCC[19, 20], breast cancer[23],
prostate cancer[22] and other tumors[24]. In prostate cancer, CENPF has been proved to be a primary regulatory factor of prostate cancer and a bleak prognostic predictor of survival and metastasis[25]. Nevertheless, the expression alteration, the molecular mechanisms and biological functions of CENPF in ACC are still unclear. This essay aimed to systematically study the expression patterns, prognosis and latent functions of CENPF in ACC.

In human ACC samples, the IHC staining showed that CENPF was significantly overexpressed and positively correlated with cell proliferation index, Ki67. It suggested that hyper-expression of CENPF may serve pivotal role in the process of cell proliferation. Compared with single array analysis, comprehensive analysis of multiple arrays has been considered as a rational and feasible method to analyze the reliability of results[34]. Thus, further insight into the expression of CENPF were gained via assaying ACC related datasets from GEO and TCGA. It confirmed that CENPF was upregulated in ACC, compared with normal ones, and positively correlated with pathological stage. The uni-cox and multi-cox analysis suggested that the CENPF expression and pTNM stage were correlated with the OS of ACC patients and might serve as independent prognostic factors in ACC patients. Thus, it suggested that CENPF could be viewed as a biomarker to make a distinction between ACC and adrenocortical adenoma or normal adrenal cortex tissues. Then, prognosis of CENPF were evaluated to announce that overexpression of CENPF were overtly related to the poor OS and PFS in ACC patients. Overexpression of CENPF were associated with poor 30-month, 60-month and 120-month OS. Our results were reinforced by numbers of publications. Huang Y et al. asserted that up-regulation of CENPF was a hazard factor for the prognosis of HCC[21]. A recent study suggested that overexpressed CENPF exerted as cancer-driver gene in the formation and development of human cancers[35]. Consequently, our results showed that the expression of CENPF was upregulated in ACC and might play a crucial role in the tumorigenesis of ACC. Then, the GSEA enrichment analysis of KEGG pathway indicated that CENPFhigh was mainly involved in cell cycle, p53 signaling pathway, and DNA replication. Moreover, the GSEA analyses of Biocarta pathway and Hallmark description uncovered that CENPF regulated the G2/M-phase cell cycle by interacting with CDK1, E2F1 and CCNB1 in human ACC.

Based on the above results, siRNA system was established to further explore potential biological functions and molecular mechanisms by performing in vitro experiments in ACC cell line, human SW13 cells. It showed that interference of CENPF inhibited the cell proliferation, adhesion between tumor cells and matrix, cell migration, cell invasion and the G2/M-phase transition, while remarkably enhanced cell apoptosis of SW13 cells. These results indicated that CENPF might induce the G2/M-phase transition by interacting with CDK1, which played a vital role in promoting cell proliferation. And, the overexpression of CENPF induced cell migration, cell invasion and inhibiting cell apoptosis. By western blotting, interference of CENPF and CDK1 by siRNA, the phosphorylation of p53 (P-p53), the expression levels of p21, and Bax were decreased compared with the siNC in SW13 cells. Therefore, the CENPF upregulated the expression level of CDK1, the phosphorylation of p53 and altered downstream protein levels, including apoptosis and cell cycle related proteins.

Previous literature reports enhanced our results. Kojima K et al. demonstrated that CDK1 inhibitor enhanced p53 mediated mitochondrial apoptosis by Bax activation and the G2/M-phase cell cycle arrest in acute myeloid leukemia[36]. Chen S et al. corroborated that the activation of p53 and p21 can inhibit the expression of CDKs, E2Fs and other factors that promote DNA replication in cell cycle of G1/S-phase arrest in bladder cancer[37]. Danupon N et al. reported that CCNB1/CDK1 complex were relocated to mitochondria during G2/M phase to phosphorylate and activate p53 at Ser-315, thus inducing anti-apoptotic response in HCT116 cells[38]. Bowen X et al. asserted that CCNB1/CDK1 regulated mitochondrial energy metabolism, promoted cell cycle progression and
tumor response to radiotherapy\[39, 40\]. Abnormal mitosis induced by CCNB1/CDK1 complex is an enormous element of cancer development or progression\[41\]. Combined with previous related research, we draw the conclusion that overexpression of CENPF upregulated CDK1 mediated G2/M-phase transition, cell proliferation, cell migration and invasion. Meanwhile, overexpression of CENPF activated p53 mediated anti-tumor effect by inducing P-p53, p21 mediated G2/M-phase cell cycle arrest or Bax mediated cell apoptosis in ACC (Figure 8).

Although a variety of biomarkers or underlying molecular mechanisms have been reported, there is still a lack of effective drugs for ACC patients\[42\]. Accordingly, it is necessary to further investigate the potential target genes or immunotherapy related to the high expression of CENPF. Firstly, immune microenvironment has been viewed as one of hallmarks of multiple cancer \[43\]. In this study, immune cell population of ACC with CENPF\textsuperscript{high} and CENPF\textsuperscript{low} was analyzed by CIBERSORTx to show that the cell subtypes were significantly different in ACC samples of CENPF\textsuperscript{high}, compared with CENPF\textsuperscript{low}. Specifically, follicular helper T cells, M0 macrophages, Eosinophils was evidently enriched in ACC with CENPF\textsuperscript{high}, compared with CENPF\textsuperscript{low}. Moreover, patients with high CENPF expression characterized higher TMB score and MSI score. Additionally, the expression of LAG3, CTLA4, HAVCR2, PD-1 and PD-L1, closely related to immunosuppression, were not evidently different in ACC patients of CENPF\textsuperscript{high} group and CENPF\textsuperscript{low} group. The result indicated that overexpression of CENPF might accumulate TMB and MSI in the process of DNA replication. Moreover, gene-drug interaction network was constructed by CTD and found that multiple CENPF inhibitors, including Cisplatin, Sunitinib, and Etoposide, might act as beneficial targets for the treatment of ACC. However, identification and verification of these assumption are still demanded further molecular mechanisms, pharmaceutical or clinical trials in the future.

Conclusion

In summary, this study pointed out that CENPF is significantly overexpressed in ACC patients. The overexpressed CENPF predicted poor prognosis of ACC and might augment the progress of ACC. The results showed that CENPF might consider as a prognostic biomarker or therapeutic target for ACC patients. According to in vitro experiments, overexpression of CENPF upregulated CDK1 mediated the G2/M-phase transition, cell proliferation, cell migration, cell invasion and accumulation of TMB and MSI. Moreover, overexpression of CENPF might activate p53 signaling pathway, which exerted potential anti-tumor effects. In brief, the study highlighted the accumulating evidence about CENPF and related signaling pathways, which might provide benefical enlightenment for the development of CENPF-mediated therapeutic drugs or formulation of strategies for ACC patients.

Materials and Methods

Clinical specimens

All Formalin-fixed and paraffin embedded (PPFE) samples were collected from patients diagnosed as ACC and adrenocortical adenoma and stored at 20~25\(^\circ\)C, from Jan 2011 to Jan 2021 at Taihe Hospital of Hubei University of Medicine, China. According to the pathological characteristics by two pathologists, all ACC patients were diagnosed and graded. Finally, 6 cases of ACC, 12 cases of adrenocortical adenoma (benign tumor) and 12 cases of normal adrenal cortex samples (adjacent to adrenocortical adenoma) were involved in this study. Details of all enrolled patients were listed in Supplementary Table 1.

Histological Analysis
Hematoxylin-eosin (HE) and immunohistochemistry (IHC) staining of PPFE samples were conducted following the steps in the manufacturer's instruction. Specifically, 3 μm of ACC tissues, adrenocortical adenoma tissues and normal adrenal cortex tissues were sliced from the PPFE. For HE, tissue sections were stained with hematoxylin-eosin staining kit (E607318-0200, Sangon, Shanghai) for morphological observations. For IHC, all slices were dewaxed with xylene and rehydrated with graded ethanol. The 3% hydrogen peroxide was used to block endogenous peroxidase activity in methanol for 10 min. The sections were incubated with CENPF primary antibody (Supplementary Table 2) for 1 h at 37 °C, and incubated with HRP labeled second antibody at 37 °C for 0.5 h, hematoxylin staining was performed at 37 °C for 30 seconds. Lastly, CENPF positive cells were tallied in 5 random high-power fields (400 ×), and the average positive cell ratio was calculated.

Bioinformatics data processing

The Gene Expression Omnibus (GEO) of NCBI, an open access genomics database, was widely used for systematic analysis of cancer genes. Only two GEO datasets were retrieved for analysis of adrenocortical carcinoma, including GSE90713 and GSE19750. Details of those datasets were listed in Supplementary Table 3. The ACC data of TCGA were derived from Genomic Data Commons Data Portal, containing 79 ACC samples.

As an interactive online service platform, Gene Expression Profiling Interactive Analysis (GEPIA) is widely conducted to analyze gene expression of tumors and control samples from the TCGA and GTEx datasets[44]. Thus, the GEPIA dataset was utilized to measure the mRNA expression of CENPF between ACC and normal tissues, and in different pathological stages.

Correlation between CENPF expression and clinicopathological parameters in ACC

The correlation between CENPF and clinicopathological features was studied based on TCGA-ACC dataset. The 79 ACC tissues in the expression matrix were divided into two groups, including 39 ACC tissues with CENPF low expression (CENPF<sub>low</sub>) and 40 ACC samples with CENPF high expression (CENPF<sub>high</sub>) by the median cutoffs. The expression groups were grouped by 'ggrisk' package of R software (version 4.0.0)[45, 46]. Sanguini diagram was drawn via the 'ggalluval' package [47] for displaying the distribution of the gene expression in survival status, ages, genders, stages and other clinical characteristics for ACC. Then, the univariate (uni-cox) and multivariate cox (multi-cox) regression analysis were analyzed and applied to develop the Nomogram. Then, 'forestplot' R packages were applied to present several parameters, including the P-values, HR and 95% confidence interval (CI) by forest. [48] Based on the results of multi-cox analysis, The Nomogram was constructed to provide a graphical representation of the risk factors and calculate the 1-year, 2-year, and 3-year overall recurrence for an ACC patient via 'rms' R package[49].

Prognosis of CENPF expression in ACC

The prognosis analysis, including overall survival (OS) and progression-free survival (PFS) were performed in ACC with CENPF<sub>high</sub> and CENPF<sub>low</sub> by 'survival' and 'survminer' packages[45]. The timeROC analysis was applied to predict accuracy of CENPF and risk score by 'timeROC' package[46], and AUC threshold is set to 0.80.

Gene-set enrichment analysis (GSEA)
The GSEA software 4.0.3 (Broad Institute, USA) was applied to probe and uncovered biological mechanisms in TCGA-ACC patients, including 39 CENPF^low and 40 CENPF^high ACC samples[50, 51]. The four predefined gene-sets, including 'c2.cp.kegg.v7.2.symbols.gmt', 'h.all.v7.2.symbols.gmt', 'c2.cp.go.v7.2.symbols.gmt', and 'c2.cp.biocarta.v7.2.symbols.gmt' were analyzed. Normalized enrichment scores (NES) were reckoned as the main GSEA statistic results. Statistical significance threshold were set as |NES| > 1, normalized P-values (NOM P-values) < 0.05 and FDR < 0.25. Additionally, analysis of protein-protein interaction (PPI) network was performed to explore the interaction between CENPF and related proteins based on The Search Tool for the Retrieval of Interacting Genes (STRING) database (version 11.5).

**Immune cell infiltration signatures**

As one of crucial indicators to speculate the effect of immunotherapy, the immune cell infiltration in tumor has become a research hotspot[52]. CIBERSORTx, the online service platform that estimated the abundance of multiple cell-types in the mixed cell population by inputting a standardized gene expression matrix[53], was carried out to analyze immune cell infiltration based on TCGA-ACC dataset. The results showed the abundance of 22 types of immune cells, including T cells, B cells, natural killer (NK) cells, monocytes, macrophages (Mφ), dendritic cells (DC) and granulocytes (mast cells, eosinophils, and neutrophils). 'Limma' package of R software was performed to normalize the expression matrix from TCGA dataset. Then, the expression of LAG3, CTLA4, PD-1, PD-L1 and HAVCR2, closely related to immunotherapy, were explored in ACC[54]. In order to predict anti-tumor effect of immunosuppressive agents, the 'ggstatsplot' package of R software were further conducted to analyze to the correlation between the expression of CENPF and microsatellite instability (MSI) or tumor mutation burden (TMB) [55].

**Gene-drug interaction network**

The gene-drug interaction network of CENPF was constructed via Comparative Toxicogenomics Database (CTD) [56] for chemotherapeutic drugs that could reduce or elevate the mRNA or protein expression levels of CENPF, and visualized by the OmicShare tools.

**Human SW13 cells in vitro assay**

The siRNA experiment in

The siRNA transfections were performed according to the manufacturer's instructions. As one of human ACC cell lines, SW13 cells were transfected with 50 nmol of CENPF siRNA (siCENPF), CDK1 siRNA (siCDK1) or negative control siRNA (siNC) in special medium (CM0451, Procell, China) for 48 h. Then, SW13 cells were lysed by TRIzol reagent (Invitrogen, USA) for total RNA isolation. The cDNA was obtained by reverse transcriptase kit (Invitrogen, USA). SYBR Green PCR Mixture (Qiagen, Germany) and specific primers were performed in ABI Prism 7500 analyzer (Applied Biosystems, USA). GAPDH was endogenous reference gene. Three replicates were set for all reactions. The $2^{-\Delta\Delta CT}$ method was applied to calculate the relative expression of CENPF or CDK1 in ACC samples. The details of primers were listed in Supplementary Table 4.

Western blotting

The human SW13 cells were transfected with siRNA targeting CENPF, CDK1 or control siRNA for 48 h. Cell lysates were denatured for SDS-PAGE electrophoresis, and then were transferred to PVDF membranes (Millipore, USA). The PVDF membranes were incubated with CENPF or CDK1 antibody (Supplementary Table 2) overnight at 4°C and incubated with the secondary HRP-labeled
antibody for 1 h at room temperature. Lastly, the PVDF membranes were visualized by Lumi-Glo reagents (Millipore, USA).

Cell proliferation assay
A total 1×10^4 cells of human SW13 cells were seeded in 96-well plates at 100 μL and transfected with 50 nmol of siCENPF, siCDK1 or siNC, then added 10 μL CCK-8 solution (Beyotime, China) to each well. After incubated for 96 h, the absorbance (optical density, OD), representing the density of cells, was measured at 450 nm.

Cell migration assay
Wound healing experiment were performed to analyze cell migration. Human SW13 cells were plated in 12-well plates in DMEM with 10% fetal bovine serum (FBS) and transfected with 50 nmol of siCENPF, siCDK1 or siNC for 48 h. A 20μl pipette tip was utilized to make wounds. Then each well was washed 5 times by PBS to remove the floating cells, and 4 mL DMEM (10% FBS, 1% antibiotic-antimycotic) was subjected. The scratch areas were drawn at 0 and 48 h.

Cell adhesion assay
Human SW13 cells were seeded in 12-well plates (1×10^5 cells/mL), transfected with 50 nmol of siRNAs for 48 h, and then transferred to 24-well plates for 3 h. All cells were rinsed and fixed with 4% PFA for 20 minutes. Then, cells were stained by Cristal-violet Staining Kits (Beyotime, China) and incubated for 10 minutes. Cells adhered to the stroma were photographed and counted by Image J software.

Cell invasion assay
Human SW13 cells transfected with siCENPF, CDK1 or siNC for 48 h were plated in TranswellR cell culture chambers (Corning, USA) with 1×10^4 cells/well. The upper chamber of a TranswellR insert was filled with cell suspension. High concentration FBS (20% FBS), as chemoattractant, was subjected to the lower chamber for 24 h. The cells under the membrane were fixed with 4% PFA and stained with crystal violet. Cells were photographed and counted by Image J software in ten random fields per chamber.

Cell apoptosis analyses
Human SW13 cells transfected with siCENPF, CDK1 or siNC for 48 h were disposed with Annexin V-FITC kit (Beyotime Biotechnology, China) and analyzed by flow cytometry (FACSCalibur, Bio-Rad, USA) to detect cell apoptosis. Data were analyzed using FlowJo7.6 software.

Cell cycle assay
For cell cycle assay, cell cycle assay kit (ab112116, Abcam) was purchased. After transfected with siCENPF, CDK1 or siNC for 48 h, SW13 cells were harvested and fixed in 70% ice-cold ethanol overnight at 4 °C. Then cells were analyzed by the Cell Cycle Assay Kit (ab112116, Abcam) in flow cytometer. Lastly, data were analyzed by FlowJo7.6 software.

Statistical analysis
Statistical analysis was conducted via SPSS 22.0 (IBM SPSS Inc. Chicago, IL) and visualized by GraphPad Prism 9.0 (San Diego, CA) software. Comparison of different groups were conducted by
two-tail unpaired Student's *t*-test or one-way analysis of variance. *P* < 0.05 was considered as a significant difference.

**Abbreviations**

ACC: Adrenocortical carcinoma; CENPF: Centromere protein F; TCGA: The Cancer Genome Atlas; OS: Overall survival; PFS: Progression-free survival; GEO: Gene expression omnibus; ROC: Receiver operator characteristic; qRT-PCR: Reverse transcription-quantitative PCR; IHC: Immunohistochemistry; CCK-8: Cell Counting Kit-8; GSEA: Gene set enrichment analysis; GO: Gene Ontology; KEGG: Kyoto Encyclopedia of Genes and Genomes; CCND1: Cyclin D1; FDR: False discovery rate; SD: Standard Deviation.

**Author Contributions**

Yu-gang Huang and Dan Li conceived, designed, performed statistical analysis and wrote the paper. Xiao-min Su and Xian-bin Tang supervised the research; Li Wang participated in the study design and provisioned useful suggestions in methodology and figure preparation. All authors have read and approved the final version of the manuscript.

**Competing Interests**

The authors have declared that there are no competing interests associated with the manuscript.

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**Availability of data and materials**

All publicly available datasets analyzed in this study can be acquired from GEO (https://www.ncbi.nlm.nih.gov/geo/) and TCGA (https://portal.gdc.cancer.gov/).

**Declarations**

**Ethics approval and consent to participate**

All clinical specimens were collected by informed consent (IFC) from patients or family members, and this study has been supported and approved by the Ethics Committee of Taihe Hospital.

**Consent for publication**

Yes.

**Competing interests**

The authors declare no conflict of interests.

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**Figure legends**

**Figure 1.** CENPF was overexpressed in ACC patients, compared with normal adrenocortical samples or adrenocortical adenoma. (A) Gross specimen photography for adrenocortical adenoma (upper) and ACC (bottom). (B) Percentage of CENPF positive cells in ACC via IHC (high power fields, 400 ×). (C) IHC for HE (upper) and CENPF protein expression (bottom) in normal adrenal cortex, adrenocortical adenoma and ACC samples (high power fields, 400 ×). The expression of CENPF was denoted by the red arrows. The expression of CENPF from (D) GSE90713 and (E) GSE19750. (F) Relative expression of CENPF in ACC, normal tissue and in tumor stage I-IV (G) based on GEPIA online server.

**Figure 2.** The expression of CENPF was positively related to the cell proliferation index, Ki67. (A) and (B) The proportion of CENPF positive cells was positively correlated with the proportion of Ki67 positive cells (n=6). (C) Based on TCGA-ACC, the expression of CENPF was positively correlated with that of MKI67 at RNA level. The expression of CENPF or Ki67 was depicted by the red arrows.

**Figure 3.** Prognosis analysis of CENPF in clinicopathological parameters for ACC patients. (A) CENPF expression (high or low) and survival status (alive or dead) of ACC patients. All the sample order is consistent. (B) Sanguini diagram for depicting the distribution of CENPF expression in age, gender, pTNM stage and survival status. The P-value and hazard ratio of CENPF and associated parameters of the uni-cox (C) and multi-cox (D) analysis. (E) Nomogram for assessing the survival rate of 1-year, 2-year, and 3-year in ACC patients related to CENPF expression.

**Figure 4.** The prognosis of CENPF expression in ACC patients. The OS (A), ROC curves of CENPF at 1-year, 3-year, and 5-year for OS (B), PFS (C), ROC curves of CENPF at 1-year, 3-year, and 5-year for PFS (D), survival
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**Figure 5.** *CENPF* regulates cell cycle by interacting with *CDK1, E2F1* and *CCNB1* in ACC. (A) GSEA analysis of KEGG pathway showed that cell cycle, DNA replication, nucleotide excision repair, and p53 signaling pathway were overtly involved in ACC with CENPF**high**. (B) GSEA analysis of Biocarta pathway and Hallmark description suggested that overexpressed *CENPF* was positively related to cell cycle (especially G2-phase of mitosis), G2/M checkpoint, and E2F targets. (C) Venn graph demonstrated that 20 genes, including *CDK1, E2F1*, and *CCNB1*, were overlapped in at least two groups. (D) Top 10 GSEA analysis of GO terms. (E) Correlation analysis between *CDK1, CCNB1, E2F1* and *CENPF* expression in ACC. (F) Expression of *CDK1* and *E2F1* in ACC and normal samples. (G) Survival analysis of *CDK1, CCNB1, and E2F1* in ACC based on GEPIA database. (H) PPI network figured out the close interactions among CENPF, CDK1 and CCNB1.

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**Figure 7.** The potential therapeutic strategies or available chemical drugs for the treatment of ACC. (A) Immune infiltration analysis of CENPF**high** (n=6) and CENPF**low** (n=7) samples in ACC. The immune fraction (B) and PCA analysis (C) in ACC samples. (D) Screening the expression level between the 5 genes, including *LAG3, CTLA4, HAVCR2, PD-1* and *PD-L1*, and CENPF in ACC. Correlation between CENPF expression and MSI(E) and TMB(F) in ACC. The horizontal axis represents the gene expression distribution, and the vertical axis denotes the expression distribution of TMB/MSI score. The density curve on the right side represents the distribution trend of TMB/MSI score; The density curve of the upper part represents the distribution trend of genes; The value represents the correlation $P$-value, correlation coefficient and calculation method. (G) The gene-drug interaction network of CENPF and related chemotherapeutics. Red arrows: chemotherapeutics up-regulate CENPF expression; green arrows: chemotherapeutics down-regulate CENPF expression. The numbers of arrows between the chemotherapeutics and CENPF in this network denote the supported numbers of previous literatures.

**Figure 8.** Mechanism diagram. Overexpressed CENPF could induce the expression of CDK1, augment G2/M-phase conversion, promote cell proliferation and lead to the accumulation of abnormal gene mutations, namely TMB, during the process of DNA replication or cell cycle. Meanwhile, these tumor signatures, including DNA damage etc., might trigger the activation of p53 signaling pathway and the downstream expression of p21 and Bax, thus inducing p53 mediated anti-tumor effect by arresting cell cycle of G2/M-phase and boosting cell apoptosis in ACC.
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**Supplementary Files**

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- SupplementaryFigure1.docx
- SupplementaryTable1.docx
