

# The genetic association between type 2 diabetic and hepatocellular carcinomas

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**Background:** Type 2 diabetes mellitus (T2DM) and hepatocellular carcinoma (HCC) are both major health problems throughout the world. It has been reported that T2DM is an independent risk factor for HCC, although the pathophysiology is still unclear.

**Methods:** In order to identify differentially expressed genes (DEGs) in T2DM and HCC, gene expression datasets for T2DM (GSE15653), HCC (GSE60502) and metformin-treated cells (GSE69850) were obtained from the Gene Expression Omnibus database repository. Protein-protein interaction (PPI) networks for the DEGs were constructed and gene clusters selected for functional enrichment analysis. Ten genes with the highest degree of connectivity were selected as hub genes and prognostic analysis together with analysis of gene expression and protein distribution were performed for these genes. Lastly, we investigated associations between the hub genes and genes associated with metformin treatment in hepatocarcinoma cells.

**Results:** In total, 256 common DEGs, including 155 up-regulated genes and 101 down-regulated genes, were identified. Enrichment analyses showed that the genes of the major module were largely associated with the cell cycle. All of the 10 hub genes (*CCNA2*, *CCNB1*, *MAD2L1*, *BU1B*, *RACGAP1*, *CHEK1*, *BUB1*, *ASPM*, *NCAPG* and *TTK*) have a strong association with lower overall survival in liver cancer patients and four genes (*CCNA2*, *CCNB1*, *CHEK1* and *BUB1*) have reduced expression in metformin-treated samples.

**Conclusions:** This study identified a number of genes that may play important roles in the association of T2DM and HCC, including four genes which may be the target of metformin treatment for diabetes and HCC. The specific mechanisms involved remain to be identified.

**Keywords:** Type 2 diabetes mellitus (T2DM); hepatocellular carcinoma (HCC); differentially expressed gene; Metformin

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#### Introduction

Globally, liver cancers rank seventh for cancer incidence while their mortality ranks fourth among all types of cancer (1). Hepatocellular carcinoma (HCC) is a malignant tumor originating from hepatocytes and it is the most common cancer in the liver. Chronic viral hepatitis and alcoholism are the two main causes of HCC. Non-alcoholic fatty liver disease (NAFLD) is also recognized as an important risk factor for HCC due to the increasing prevalence of obesity and Type 2 diabetes mellitus (T2DM) (2). T2DM is a disease of the endocrine system and is common, with one T2DM patient per every 11 adults (3). Recently, there have been a number of reports researches showing that T2DM has a negative influence on the survival of HCC patients (4,5).

Metformin is frequently used to treat hyperglycemia associated with T2DM and has also been reported to have anti-hepatocellular carcinoma effects (6). However, most of these studies have focused on epidemiology with only a few concentrating on possible genetic connections between HCC and T2DM.

We, therefore, undertook this study to explore the specific mechanism of the association between T2DM and HCC. We analyzed the gene expression profiles of T2DM and HCC tissue to identify differentially expressed genes (DEGs) in both diseases, followed by the construction of protein-protein interaction (PPI) networks for genes observed to be up-regulated or down-regulated in both T2DM and HCC. After construction of the PPI network, a gene cluster would be selected for functional enrichment analyses. Ten genes with the highest degree of connectivity were selected as hub genes. Prognostic analysis, gene expression analysis and protein distribution analysis were performed on the hub genes. We also investigated associations between the hub genes and genes implicated in metformin treatment for HCC. Lastly, we discuss our results in relation to published findings in an attempt to elucidate the relationship between T2DM and HCC.

## **Methods**

# Microarray data

In order to avoid errors caused by platform differences, the datasets containing T2DM samples and HCC samples should use the same platform. Two gene expression datasets GSE15653 (GPL96; T2DM) and GSE60502 (GPL96;

HCC) were chosen from the Gene Expression Omnibus (GEO, https://www.ncbi.nlm.nih.gov/geo/) database repository. In addition, the dataset GSE69850 (GPL13667), a HepG2 hepatocarcinoma cell dataset which contains metformin-treated samples, was also downloaded from the GEO (*Table 1*).

## Identification of DEGs

We used GRO2R (https://www.ncbi.nlm.nih.gov/geo/geo2r/), a web tool that allows users to identify DEGs between two or more groups of samples in a GEO series, analyze gene expression in GSE15653 and GSE60502. IlogFC | ≥1.0 and P<0.05 were the criteria used for gene selection as DEGs and statistical significance, respectively. Genes showing up-regulation or down-regulation in both T2DM and HCC were investigated by Venn diagram, using the Venn diagram web tool (http://bioinformatics.psb. ugent.be/webtools/Venn/).

## PPI network construction

We used STRING (https://string-db.org), a database of known and predicted protein-protein interactions, to construct the PPI network for genes that were upregulated or down-regulated in both T2DM and HCC. Interactions with a score >0.7 were considered statistically significant. We visualized the PPI networks by Cytoscape, an open-source software platform for visualizing molecular interaction networks and biological pathways. Ten genes with the highest degree of connectivity were selected as hub genes. The MCODE plugin in Cytoscape was used to find clusters with the degree cut-off, haircut on, k-core, node score cut-off, max depth set as 10, 0.2, 2, 0.2 and 100 in PPI network. Modules with the highest scores were selected for functional enrichment analyses.

# Functional enrichment analyses for module genes

The Kyoto Encyclopedia of Genes and Genomes (KEGG) is a database resource for understanding high-level functions of biological systems from molecular-level information. The Gene Ontology (GO) resource can be used to perform enrichment analysis. We used DAVID (https://david.ncifcrf.gov/) for both KEGG pathway and GO enrichment analysis, including Biological Process, Cellular Component and Molecular Function for the selected module genes.

Table 1 Description of data used in this study

Accession	Platform	Normal/control	Cancer/treat	Type
GSE15653	GPL96	5 normal	4 diabetes	Liver tissues
GSE60502	GPL96	18 normal	18 cancer	HCC tissues
GSE69850	GPL13667	39 control	9 Metformin treated	HepG2 cells

HCC, hepatocellular carcinoma.

## Analysis of hub genes

To identify the prognostic values of the hub genes, we used Kaplan-Meier Plotter database (https://kmplot.com) to construct Kaplan-Meier survival plots for 10 hub genes. GEPIA (http://gepia.cancer-pku.cn), a web server for cancer and normal gene expression profiling and interactive analysis, was used to analyze the expression of hub genes in HCC and normal tissue and to verify the prognostic values of hub genes in the Cancer Genome Atlas (TCGA) database. Lastly, the Human Protein Atlas (https://www.proteinatlas.org) was used to analyze the protein distribution of hub genes in HCC and normal liver tissue.

# Analysis of genes associated with metformin treatment

Unpaired *t*-tests were used in GraphPad Prism to analyze the expression of 10 hub genes in GSE69850, which contains nine samples of HepG2 hepatocarcinoma cells treated with metformin and 39 control samples treated with dimethyl sulfoxide (DMSO), in order to determine associations between the hub genes and metformin treatment for HCC. Results with P<0.05 were considered statistically significant (*Figure 1*).

## **Results**

## Identification of DEGs in HCC and T2DM

In total, 2,462 DGEs were identified from four T2DM samples and five control samples in dataset GSE15653 and 1,760 DGEs were identified from 18 HCC samples and 18 adjacent non-tumorous liver samples in dataset GSE60502. Then, 256 genes were found to overlap between the T2DM and HCC DGEs, including 155 up-regulated genes and 101 down-regulated genes (*Figure 2*).

## PPI network construction

PPI network maps yielded 223 nodes and 531 edges, indicating the interactions between the 256 overlapping genes. The ten hub genes with the highest degree of connectivity were CCNA2, CCNB1, MAD2L1, BUB1B, RACGAP1, CHEK1, BUB1, ASPM, NCAPG, TTK (Figure 3A, Table 2). We use the MCODE plugin in Cytoscape to analyze the PPI networks and choose the module which has a highest score to create a cluster network including 31 nodes and 433 edges (Figure 3B).

## Functional enrichment analyses for module genes

We used DAVID to make functional enrichment analyses for the 31 module genes selected by MCODE. The major biological processes identified for these genes are "mitotic sister chromatid segregation" and "mitotic cell cycle phase transition" (*Figure 4A*). Cellular component analysis indicated that these genes are mainly enriched in "spindle" and "nuclear chromosome part" (*Figure 4B*). In terms of molecular function, these genes are mainly associated with "microtubule-binding" and "tubulin-binding" (*Figure 4C*). Lastly, through KEGG pathway analysis, we observed that these genes are mostly enriched in "cell cycle" (*Figure 4D*) (*Tables 3-6*).

# Hub genes analysis

The results of the prognostic analysis showed that all of the 10 hub genes (*CCNA2*, *CCNB1*, *MAD2L1*, *BU1B*, *RACGAP1*, *CHEK1*, *BUB1*, *ASPM*, *NCAPG* and *TTK*) have strong associations with a lower overall survival rate in liver cancer patients (*P*<0.05) (*Figure 5*). Likewise, GEPIA analysis showed that the 10 hub genes are all overexpressed

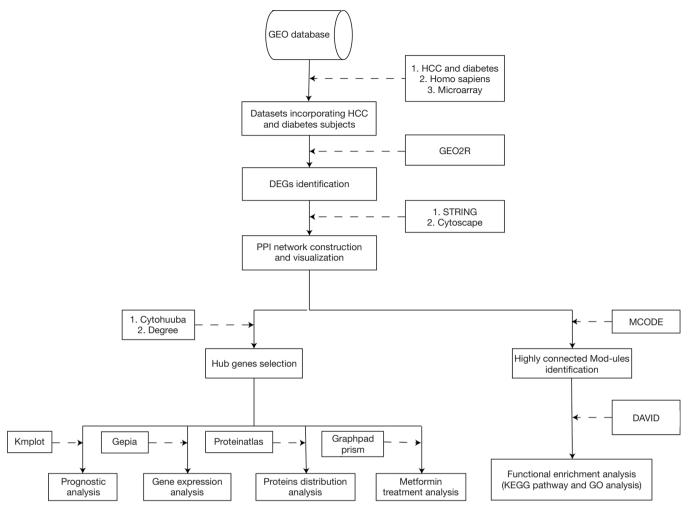


Figure 1 Graphical abstract.

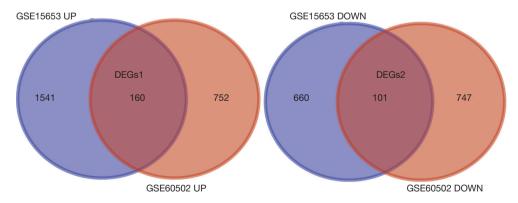


Figure 2 Differentially expressed genes among diabetes DEGs and HCC DEGs. DEGs1 represents the genes that up-regulated in both diabetes and HCC. DEGs2 represents the genes that down-regulated in both diabetes and HCC. DEGs, differentially expressed genes; HCC, hepatocellular carcinoma.

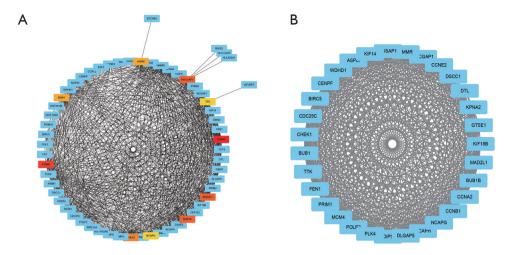


Figure 3 PPI network construction and module analysis on the DEGs1 and DEGs2. (A) PPI network on the DEGs1 and DEGs2, the red, orange and yellow nodes represent hub genes; (B) the chosen module which has a highest score using MCODE. DEGs, differentially expressed genes; PPI, protein-protein interaction.

Table 2 Top 10 hub genes

Rank	Name
1	CCNA2
2	CCNB1
3	MAD2L1
3	BUB1B
3	RACGAP1
6	CHEK1
7	BUB1
7	ASPM
9	NCAPG
9	TTK

in HCC (*Figure 6*). Lastly, using The Human Protein Atlas, we investigated the protein distribution data of six genes: *CCNA2*, *CCNB1*, *MAD2L1*, *RACGAP1*, *NCAPG*, *TTK*. All were found to be overexpressed in HCC (*Figure 7*).

## Association between bub genes and metformin treatment

The GSE69850 dataset contains nine HepG2 samples treated by metformin and 39 controlled samples treated by DMSO. After analyzing the expression of the 10 hub genes in metformin-treated and control samples, we found that four genes, *CCNA2*, *CCNB1*, *CHEK1* and *BUB1*, have a

significantly lower expression in metformin-treated samples (P<0.05) (*Figure 8*), suggesting that these genes may be involved in the mechanism of metformin treatment for T2DM and HCC.

## **Discussion**

Both T2DM and HCC are globally epidemic diseases that place a huge burden on the patients and health services throughout the world. Many reports have indicated that T2DM is an independent risk factor for HCC. In this study, we have identified 256 genes that overlapped between T2DM DGEs and HCC DGEs, including 155 up-regulated genes and 101 down-regulated genes. Enrichment analyses show that the genes of the major module are largely associated with the cell cycle. Considering the ten identified hub genes, Cyclin A2, encoded by the cyclin A2 (CCNA2) gene, is a regulator of cell division and which activates Cdk1 and Cdk2, playing an important role in G1/S transition (7). Several studies have found that high levels of Cyclin A2 may be associated with carcinogenesis (8,9). It has also been reported that CCNA2 affects  $\beta$ -Cell proliferation (10) in the pancreas. Cyclin B1 (CCNB1) is a mitotic regulatory protein which complexes with CDK1 (CDC2) to form the maturation phase-promoting factor (MPF) and induces the G2/M phase transition in G2 phase cells (11). A metaanalysis has suggested that elevated expression of cyclin B1 may be associated with poor outcomes in lung cancer and esophageal carcinoma (12). It has also been found that

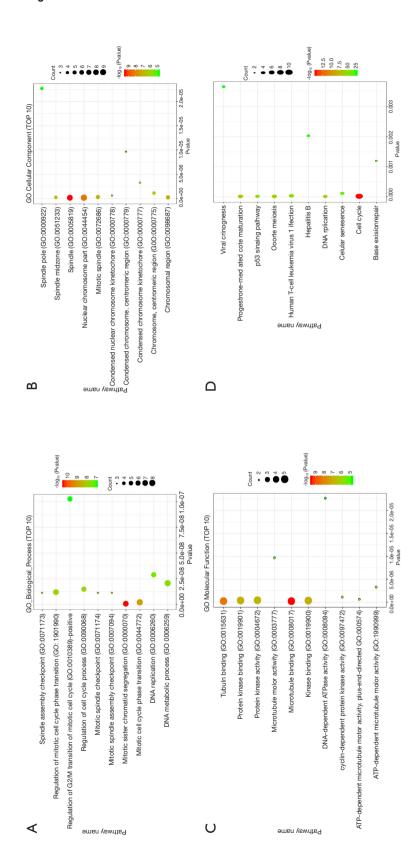


Figure 4 Functional enrichment analysis for module genes. (A,B,C,D) Represent the results of biological process, cellular component, molecular function and KEGG pathway analyses for module genes.

Table 3 KEGG pathway analysis

Term	P value	Genes
Cell cycle	2.30E-15	CCNA2; CCNB1; CCNE2; CHEK1; BUB1B; MCM4; TTK; CDC25C; BUB1; MAD2L1
DNA replication	2.69E-07	FEN1; PRIM1; POLE2; MCM4
Progesterone-mediated oocyte maturation	4.12E-07	CCNA2; CCNB1; CDC25C; BUB1; MAD2L1
Oocyte meiosis	1.31E-06	CCNB1; CCNE2; CDC25C; BUB1; MAD2L1
p53 signaling pathway	4.51E-06	CCNB1; CCNE2; CHEK1; GTSE1
Human T-cell leukemia virus 1 infection	2.03E-05	CCNA2; CCNE2; CHEK1; BUB1B; MAD2L1
Cellular senescence	1.05E-04	CCNA2; CCNB1; CCNE2; CHEK1
Base excision repair	0.0011914	FEN1; POLE2
Hepatitis B	0.0020202	CCNA2; CCNE2; BIRC5
Viral carcinogenesis 0.00		CCNA2; CCNE2; CHEK1

Table 4 GO biological process analysis

Table 1 de blological process analysis		
Term	P value	Genes
Mitotic sister chromatid segregation (GO:0000070)	3.97E-11	CCNB1; KIF18B; KIF14; NUSAP1; NCAPG; NCAPH; DLGAP5
Mitotic cell cycle phase transition (GO:0044772)	1.29E-09	PLK4; CCNA2; CCNB1; CCNE2; PRIM1; POLE2; MCM4; CDC25C
Spindle assembly checkpoint (GO:0071173)	1.11E-08	CENPF; BUB1B; TTK; BUB1
Mitotic spindle assembly checkpoint (GO:0007094)	1.11E-08	CENPF; BUB1B; TTK; BUB1
Mitotic spindle checkpoint (GO:0071174)	1.11E-08	CENPF; BUB1B; TTK; BUB1
regulation of mitotic cell cycle phase transition (GO:1901990)	1.13E-08	PLK4; CCNB1; CENPF; KIF14; BUB1B; HMMR; MAD2L1
Positive regulation of cell cycle process (GO:0090068)	1.42E-08	FEN1; CCNB1; RACGAP1; KIF14; NUSAP1; MAD2L1
DNA metabolic process (GO:0006259)	1.99E-08	WDHD1; FEN1; PRIM1; POLE2; CHEK1; MCM4; KPNA2; CDC25C
DNA replication (GO:0006260)	2.82E-08	WDHD1; FEN1; POLE2; CHEK1; MCM4; CDC25C
Regulation of G2/M transition of mitotic cell cycle (GO:0010389)	1.02E-07	PLK4; CCNB1; CENPF; KIF14; HMMR; DTL

increased cyclin B1 may promote the proliferation of HCC cells (13) and that it may be a biomarker and therapeutic target for recurrent HBV-related HCC (14). Mitotic arrest deficient 2 protein (MAD2L1) is an essential part of the mitotic spindle assembly checkpoint that ensures the proper alignment of chromosomes, strongly suggesting that MAD2L1 has a function in tumor suppression (15). Many studies have shown that the overexpression of MAD2L1 promotes the proliferation and metastasis of HCC (16). In addition, interactions between MAD2L1 and the insulin receptor have been reported, suggesting that MAD2L1

may influence the progression of diabetes (17,18). Genes budding uninhibited by benzimidazoles homolog 1 (*BUB1*) and budding uninhibited by benzimidazoles 1 homolog beta (*BUB1B*) encode serine/threonine protein kinases that play important roles in mitosis and are also essential components of the spindle checkpoint (19), with most studies suggesting both *BUB1* and *BUB1B* may contribute to the proliferation and metastasis of some types of cancer including HCC (20-22). Gene Rac GTPase activating protein 1 (*RACGAP1*) encodes a GTPase-activating protein (GAP) which is an essential component of the central spindle complex and

Table 5 GO cellular component analysis

Term	P value	Genes
Spindle (GO:0005819)	3.30E-10	CCNB1; CENPF; RACGAP1; KIF14; NUSAP1; BUB1B; TTK; MAD2L1
Nuclear chromosome part (GO:0044454)	5.48E-09	WDHD1; FEN1; CCNB1; PRIM1; POLE2; BUB1B; MCM4; BUB1; NCAPH
Chromosomal region (GO:0098687)	1.16E-07	CENPF; DSCC1; CHEK1; BIRC5; OIP5
Spindle midzone (GO:0051233)	1.44E-07	KIF18B; RACGAP1; KIF14; BUB1B
Mitotic spindle (GO:0072686)	1.92E-07	ASPM; KIF18B; RACGAP1; NUSAP1; MAD2L1
Condensed nuclear chromosome kinetochore (GO:0000778)	5.52E-07	CCNB1; BUB1B; BUB1
Chromosome, centromeric region (GO:0000775)	1.03E-06	CENPF; DSCC1; BIRC5; OIP5
Condensed chromosome kinetochore (GO:0000777)	3.21E-06	BUB1B; BIRC5; BUB1
Condensed chromosome, centromeric region (GO:0000779)	9.62E-06	BUB1B; BIRC5; BUB1
Spindle pole (GO:0000922)	2.26E-05	ASPM; CCNB1; CENPF; DLGAP5

Table 6 GO molecular function analysis

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Term	P value	Genes
Microtubule binding (GO:0008017)	1.19E-05	KIF18B; RACGAP1; KIF14; NUSAP1; BIRC5
Tubulin binding (GO:0015631)	4.28E-05	KIF18B; RACGAP1; KIF14; NUSAP1; BIRC5
Kinase binding (GO:0019900)	4.27E-04	CCNA2; CCNB1; RACGAP1; KIF14; CDC25C
Protein kinase binding (GO:0019901)	9.17E-04	CCNA2; CCNB1; RACGAP1; KIF14; CDC25C
Protein kinase activity (GO:0004672)	0.0010748	PLK4; CCNA2; CHEK1; BUB1B; BUB1
ATP-dependent microtubule motor activity, plus-end-directed (GO:0008574)	0.0011202	KIF18B; KIF14
cyclin-dependent protein kinase activity (GO:0097472)	0.0013400	CCNA2; CCNB1
ATP-dependent microtubule motor activity (GO:1990939)	0.0026130	KIF18B; KIF14
Microtubule motor activity (GO:0003777)	0.0040194	KIF18B; KIF14
DNA-dependent ATPase activity (GO:0008094)	0.0057059	DSCC1; MCM4

plays an important role in cytokinesis its GAP activity (23). *RACGAP1* can maintain cell survival (24) and has been reported to be associated with several cancers including HCC, particularly with the migration and invasion process (25-28). Gene non-SMC condensin I complex subunit G (*NCAPG*) encodes a subunit of the condensin complex, which plays a central role in mitotic chromosome assembly and segregation (29). Recently, some studies have suggested that *NCAPG* is overexpressed in HCC and suppression of *NCAPG* inhibits proliferation and induces apoptosis in HCC cells (30-32), while normal hepatocytes show

low expression of *NCAPG* (31). *CHEK1* encodes a serine/ threonine kinase, which is required for checkpoint-mediated cell cycle arrest in response to DNA damage or the presence of unreplicated DNA. It is reported that *CHEK1* plays an essential role in cell cycle progression and survival whether in the presence or absence of DNA damage (33). *CHEK1* is reported to be significantly overexpressed in HCC (34), which may be the result of reduced miR-497 in HCC (35). Interestingly, another study has suggested that *CHEK1* may induce the overexpression of *CCNB1* to promote tumor growth in human colorectal cancer (36), while a *CHEK1* 

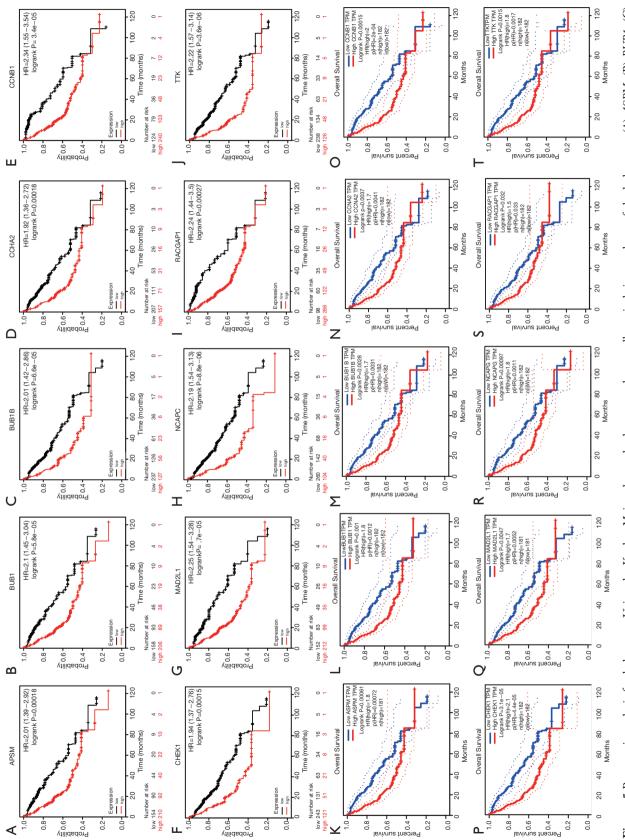


Figure 5 Prognostic analysis for hub genes. Using the Kaplan-Meier plotter database to make overall survival time analyses of the hub genes. (A) ASPM; (B) BUB1; (C) BUBIB; (D) CCNA2; (E) CCNB1; (F) CHEK1; (G) MAD2L1; (H) NCAPG; (I) RACGAP1; (J) TTK; (A-J) is in kmplot database. (K) APSM; (L) BUB1; (M) BUB1B; (N) CCNA2; (O) CCNB1; (P) CHEK1; (Q) MAD2L1; (R) NCAPG; (S) RACGAP1; (T) TTK; (K-T) is in TCGA database.

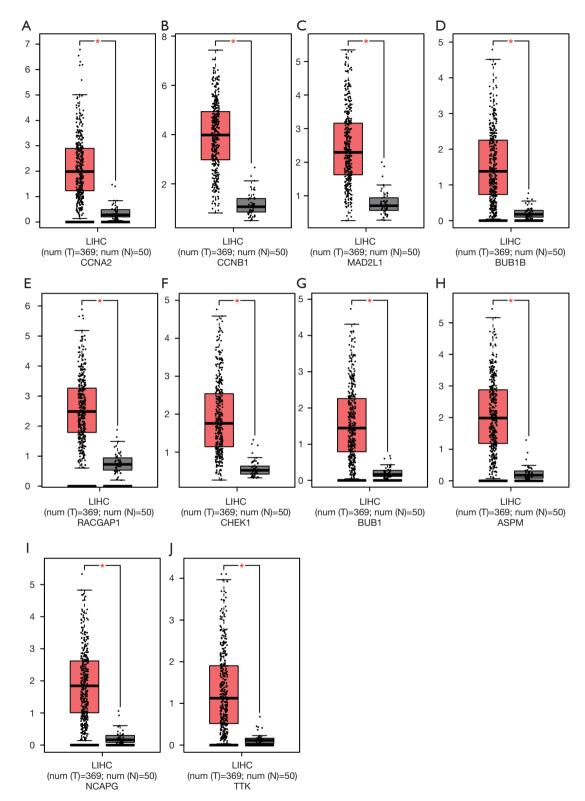


Figure 6 mRNA expression of hub genes. mRNA expression of hub genes in HCC tissue (red; n=369) and normal tissues (black; n=50). (A) CCNA2; (B) CCNB1; (C) MAD2L1; (D) BUB1B; (E) RACGAP1; (F) CHEK1; (G) BUB1; (H) ASPM; (I) NCAPG; (J) TTK. \* represents P value <0.05.

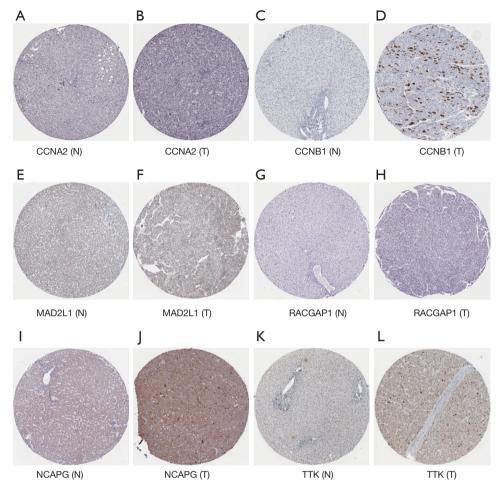


Figure 7 Protein expression of CCNA2, CCNB1, MAD2L1, RACGAP1, NCAPG, TTK. The tissue microarray using paired tumor tissue and adjacent normal tissue from the margin of the tumor. Immunohistochemistry of tissue microarray was performed by using the antibody (CAB000114, CAB003804, HPA003348, HPA043912, HPA039613, CAB013229, Abcam). (A) Protein expression of CCNA2 in normal tissue (not detected); (B) protein expression of CCNA2 in tumor tissue (staining: medium; intensity: strong; quantity: <25%); (C) protein expression of CCNB1 in normal tissue (not detected); (D) protein expression of CCNB1 in tumor tissue (staining: medium; intensity: strong; quantity: <25%); (E) protein expression of MAD2L1 in normal tissue (not detected); (F) protein expression of MAD2L1 in tumor tissue (staining: low; intensity: weak; quantity: 75–25%); (G) protein expression of RACGAP1 in normal tissue (not detected); (H) protein expression of RACGAP1 in tumor tissue (staining: medium; intensity: moderate; quantity: >75%); (K) protein expression of TTK in normal tissue(not detected); (L) protein expression of TTK in tumor tissue (staining: low; intensity: moderate; quantity: <25%).

inhibitor has been observed to have a significant effect in treating tumors (37). Abnormal spindle microtubule assembly (ASPM), a key gene for microcephaly (38), plays an essential role in regulating spindle poles (39). Several studies have suggested that ASPM is closely related to HCC and can be used as a marker for HCC metastasis and early recurrence (40,41). Gene TTK encodes a dual-specificity protein kinase with the ability to phosphorylate

tyrosine, serine and threonine. It is an essential component of the spindle assembly checkpoint and is commonly overexpressed in many human tumors (42). *TTK* has been shown to have significant effects on HCC proliferation and sorafenib resistance, and it can serve as a biomarker and potential target for HCC treatment (43,44). We found that all ten hub genes are overexpressed in cancers, including HCC, and most of them play an important role in cell

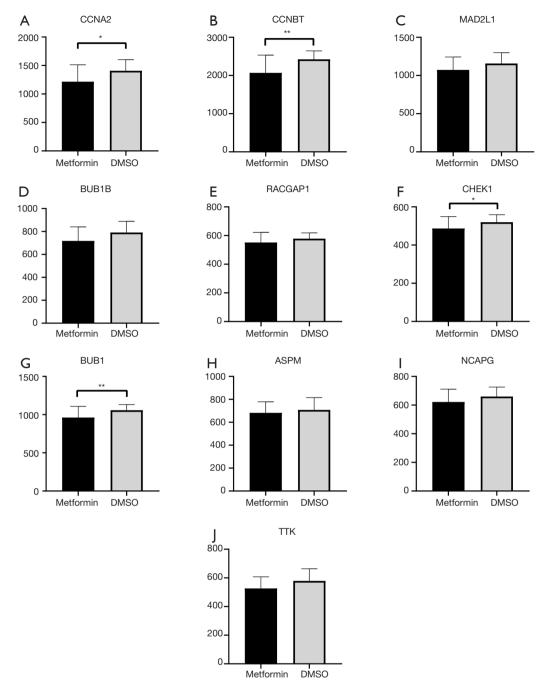


Figure 8 Changes of gene expression of HepG2 cell lines after Metformin treatment. (A-J) Represent the gene expression in 9 Metformin-treated HepG2 cell lines (black) and 39 controlled HepG2 cell lines (grey). (A) CCNA2; (B) CCNB1; (C) MAD2L1; (D) BUB1B; (E) RACGAP1; (F) CHEK1; (G) BUB1; (H) ASPM; (I) NCAPG; (J) TTK. \* represents P value <0.05; \*\* represents P value <0.01.

division. However, there are few reports of an association between these genes and T2DM.

Metformin is a commonly used hyperglycemic drug. It has been reported to have inhibiting effect on many types

of tumors including colorectal cancer, prostate cancer (45), lung cancer (46), pancreatic cancer (47) and HCC (6). It appears that its anti-tumorigenic effect may be related to the activation of AMP-activated protein kinase (*AMPK*) (48).

In this study, we found that four of the hub genes, *CCNA2*, *CCNB1*, *CHEK1* and *BUB1*, may participate in the metformin treatment mechanism for T2DM and HCC and that these genes are all enriched in cell cycle control.

It is possible that T2DM could promote the genesis of HCC via insulin resistance (49), over-activation of mTOR and chronic inflammation caused by adipokines (50). Besides, T2DM can also lead to NAFLD and non-alcoholic steatohepatitis (NASH), which are the main causes of HCC (51). Very few studies discuss the specific mechanism of the correlation between T2DM and HCC. Li *et al.* has suggested that high glucose promotes HCC cell proliferation and migration through the MEG3-MiR-483-3p-ERp29 pathway (52). The current study is the first exploration of the role of these 10 genes implicated in the combination of HCC and T2DM together with genes associated with metformin treatment for T2DM and HCC using a bioinformatics analysis.

This study has several limitations. Firstly, the sample sizes of the datasets we used are small and, secondly, our findings still need to be verified in clinical practice.

In conclusion, we have identified 256 DEGs and selected 10 genes (CCNA2, CCNB1, MAD2L1, BUB1B, RACGAP1, CHEK1, BUB1, ASPM, NCAPG and TTK) which may function biomarkers in both HCC and T2DM. We found four genes (CCNA2, CCNB1, CHEK1 and BUB) which may be the target of metformin treatment for T2DM and HCC. The specific mechanisms of action of these genes, however, require further investigation.

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#### **Footnote**

Conflicts of Interest: QX serves as an unpaid section editor of Annals of Translational Medicine from Oct 2019 to Sep 2020. The other authors have no conflicts of interest to declare.

Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved.

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