Bioinformatics analysis identified *TCP1* and *NOTCH1* as potential target molecules to overcome 5-fluorouracil resistance in cholangiocarcinoma

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Abstract

Background. Late diagnosis and chemotherapy resistance, particularly to 5-fluorouracil (5-FU), contribute to the low survival rate in cholangiocarcinoma (CCA) patients. Identifying relevant genes and pathways, as well as novel targeted molecules, is crucial to overcoming 5-FU resistance and improving treatment outcomes for CCA patients.

Objectives. This study aimed to determine the potential molecules associated with 5-FU resistance in CCA cells.

Materials and methods. Transcriptomic datasets from 4 stable 5-FU-resistant cell lines and their corresponding parental lines were retrieved from the Gene Expression Omnibus. A series of bioinformatics analyses were conducted to identify key genes upregulated in 5-FU-resistant cells compared to their parental counterparts. The expression levels of candidate genes identified through bioinformatics analysis were validated in CCA tissues and cell lines.

Results. Differential gene expression, protein–protein interaction, and Hub genes analysis revealed 8 genes that were significantly upregulated in 5-FU resistance cells compared to their parental cells. Six of the 8 genes, including *TCP1*, *RPS6*, *RPS29*, *HSPA5*, *RPS15A*, and *NOTCH1*, were upregulated in patient CCA tissues. Using real-time PCR, only the expression levels of *NOTCH1* and *TCP1* were significantly higher in the 5-FU insensitive CCA cell lines, KKU-213A and KKU-213B, than that of the 5-FU sensitive CCA cell line, KKU-055. A similar result was observed in stable 5-FU-resistant cell lines (KKU-213A-FR and KKU-213B-FR) compared to their parental cells.

Conclusions. The bioinformatic analysis and PCR results revealed that *NOTCH1* and *TCP1* might be associated with 5-FU resistance and serve as potential molecular targets to enhance 5-FU sensitivity in CCA cells.

Key words: 5-fluorouracil, resistance, cholangiocarcinoma, *NOTCH1*, *TCP1*

Highlights

- This study identifies *TCP1* and *NOTCH1* as key genes associated with 5-FU resistance in cholangiocarcinoma (CCA).
- Bioinformatics and real-time PCR analyses confirm the upregulation of *TCP1* and *NOTCH1* in 5-FU-resistant CCA cell lines and patient tissues.
- Targeting *TCP1* and *NOTCH1* may enhance the efficacy of 5-FU treatment, improving therapeutic outcomes for CCA patients.

Background

Cholangiocarcinoma (CCA) is a malignancy arising from the biliary epithelium.[1] Surgical resection remains the cornerstone of therapy for early-stage disease, offering the best chance for long-term survival. In contrast, patients with metastatic or unresectable CCA are typically managed with systemic chemotherapy, most often 5-fluorouracil (5-FU).[2] Unfortunately, response rates to 5-FU in CCA are modest, and the emergence of 5-FU resistance is the principal cause of therapeutic failure, driving disease progression and mortality. Although the molecular mechanisms underpinning 5-FU resistance have been well characterized in other cancers[3,4] those specific to CCA remain poorly understood. A deeper elucidation of these pathways is therefore essential to develop more effective, resistance-overcoming treatment strategies for CCA. Recently, bioinformatics analyses leveraging public transcriptomic repositories, such as the Gene Expression Omnibus (GEO)[5] and The Cancer Genome Atlas (TCGA)[6], have become increasingly prevalent across biomedical research. These approaches have enabled the identification of novel therapeutic targets in gastric cancer,[7] non–small cell lung cancer and esophageal carcinoma,[8] as well as cholangiocarcinoma.[9] Moreover, numerous studies have demonstrated that mining these datasets can reveal molecular drivers of anticancer drug resistance. Accordingly, applying bioinformatics methods to publicly available transcriptomic data represents a promising strategy for uncovering candidate mediators of 5-FU resistance in CCA.

In this study, we combined bioinformatics and experimental approaches to uncover molecules linked to 5-FU resistance in CCA. First, we mined multiple GEO transcriptomic datasets to identify genes consistently associated with 5-FU resistance. Next, we constructed a protein–protein interaction network (PPI) of these candidate genes to highlight central "hub" factors. We then validated the expression of key resistance-associated genes in both CCA patient tissues and established cell lines using quantitative PCR. Finally, we employed PanDrugs to predict existing compounds that target these hub genes, laying the groundwork for potential therapeutic interventions.

Objectives

This study aimed to investigate potential target molecules associated with 5-FU resistance in CCA using bioinformatics, and PCR techniques.

Materials and methods

Cholangiocarcinoma cells

Three human CCA cell lines, KKU-055, KKU-213A, and KKU-213B, were obtained from the Japanese Collection of Research Bioresources Cell Bank.[10] Cells were maintained in Dulbecco's Modified Eagle's Medium (DMEM; Gibco/BRL, Grand Island, NY, USA) supplemented with 10% fetal bovine serum and 100 U/mL penicillin–streptomycin at 37°C in a humidified atmosphere of 5% CO_2. Two 5-FU-resistant sublines (KKU-213A-FR and KKU-213B-FR), kindly provided by Assoc. Prof. S. Obchoei,[11] were cultured in complete DMEM containing the IC_{10} concentration of 5-FU.

Proliferation assay

To assess 5-FU sensitivity, 2,000 cells per well were seeded in 96-well plates and incubated under standard culture conditions. Cell viability was then quantified using the MTT assay: After treatment, MTT solution (0.5 mg/mL) was added to each well and plates were incubated for 4 h at 37°C. The resulting formazan crystals were solubilized by adding 100 μL of DMSO and mixing thoroughly. Absorbance was measured at 540 nm on a microplate reader (Tecan Austria GmbH, Salzburg, Austria) to determine relative cell viability.

Determination of differentially expressed genes

To identify molecules commonly associated with 5-FU resistance, we retrieved transcriptomic profiles of stable 5-FU-resistant cancer cell lines from the GEO database (accession numbers GSE196900, GSE23776, and GSE81005). We then examined the GSE7631 dataset – which comprises 92 non-tumor controls and 91 Thai CCA patient tissue

samples, to evaluate the expression of candidate genes in clinical specimens.

Differentially expressed genes (DEGs) were identified from microarray datasets using GEO2R (https://www.ncbi.nlm.nih.gov/geo/geo2r) and from RNA-seq data via the Galaxy platform (https://usegalaxy.org).[12] Genes meeting the criteria of an adjusted p-value < 0.05 and |log$_2$ fold change| > 0.5 were considered significant. Data visualization, including volcano plots and Venn diagrams, was performed in RStudio (https://rstudio.com) and with the jvenn online tool (https://jvenn.toulouse.inrae.fr/app/index.html), respectively.

Gene Ontology and pathway enrichment analysis

Gene Ontology (GO) and pathway enrichment analyses of the identified DEGs were performed using the DAVID Functional Annotation tool.[13] Differentially expressed genes were classified into GO categories, biological processes, cellular components, and molecular functions, as well as mapped to Reactome pathways.[14] Enrichment results were visualized as a bubble plot in RStudio to illustrate the significance and gene counts for each term.

Protein–protein interaction and hub gene identifications

Protein–protein interaction (PPI) networks for the common DEGs were generated using STRING v. 11.5 (https://string-db.org/). Hub genes within these networks were identified and ranked in Cytoscape (https://cytoscape.org) via the cytoHubba plugin,[15] with the top 30 genes selected based on connectivity degree. A heatmap illustrating the expression patterns of these hub genes across all datasets was then plotted using GraphPad Prism 9.2 (GraphPad Software, San Diego, USA).

Real-time polymerase chain reaction

Total RNA was isolated from CCA cell lines using TRIzol® Reagent (Invitrogen, Carlsbad, USA) according to the manufacturer's protocol. RNA purity and concentration were assessed with a NanoDrop™ 2000 Spectrophotometer (Thermo Fisher Scientific, Waltham,

USA). For reverse transcription, total RNA was converted to cDNA using the High-Capacity cDNA Reverse Transcription Kit (Applied Biosystems, Foster City, USA; cat. No. 4368814) following the supplier's instructions.

Real-time quantitative PCR was conducted to assess the mRNA expression of candidate 5-FU resistance genes in CCA cell lines using a LightCycler® 480 system (Roche Diagnostics, Basel, Switzerland). Each 20 µL reaction contained 1× LightCycler® 480 SYBR Green I Master Mix (Roche Diagnostics), 0.5 µM of each primer, and 40 ng of cDNA template. Thermal cycling and fluorescence acquisition were performed according to the manufacturer's recommended protocol.

PCR amplifications were performed on a LightCycler 480 system with the following cycling conditions: An initial denaturation at 95°C for 5 min; 40 cycles of 95°C for 20 s, gene-specific annealing at 65°C (*TCP1*), 58°C (*RPS6*, *RPS15A*, *NOTCH1*), 60°C (*RPS29*), or 63°C (*HSPA5*) for 10 s; and extension at 72°C for 20 s. For each reaction, cycle threshold (Ct) and melting temperature (Tm) values were recorded, and mean ±SD were calculated. Relative gene expression was quantified using the $2^{-\Delta Ct}$ method. Primer sequences and expected amplicon sizes are detailed in Table 1.

Determination of targeted drug for 5-FU sensitizing using PanDrugs analysis

PanDrugs (www.pandrugs.org)[16] was used to identify druggable targets for sensitizing 5-FU-resistant CCA cell lines. Hub genes from the PPI analysis were submitted to PanDrugs, and candidate compounds were selected based on 2 criteria: 1) availability of a clinically approved, specific inhibitor targeting the gene product, and 2) evidence of the drug's efficacy in enhancing chemosensitivity.

Statistical analyses

Real-time PCR data are reported as the mean ±SD from 3 independent biological experiments, each performed in triplicate. Statistical analyses were conducted using IBM SPSS v. 24.0 (IBM Corp., Armonk, USA) and GraphPad Prism 9.2 (GraphPad Software Inc.). For microarray and RNA-seq datasets, p-values were calculated using the Limma and DESeq2 packages in R, respectively,

Table 1. The list of primers for candidate 5-FU associated genes

Genes	Forward primers	Reverse primers	Product sizes (bp)
TCP1	CGACTTCTGCCCATTCTCTC	CTCTTGGAGCATCTGGCTGT	70
RPS6	TGCTCTGAAGAAGCAGCGTA	GGAAAGTCTGCGTCTCTTCG	130
RPS29	TGGCTCTAGAAGTGGCTGGT	GTGAAGGCAAGGTTGGTCAT	110
HSPA5	TAGCGTATGGTGCTGCTGTC	TGACACCTCCCACAGTTTCA	117
RPS15A	CCTGTCCCCTAGTCTCTGCT	GAAGCTGATCCATGCCCCTT	93
NOTCH1	GGAGGCATCCTACCCTTTTC	TGTGTTGCTGGAGCATCTTC	118

5-FU – 5-fluorouracil; bp – base pair.

and adjusted for multiple testing by the Benjamini–Hochberg method. Normality of data distributions was assessed by the Kolmogorov–Smirnov test, and homogeneity of variances was confirmed by Levene's test (Supplementary Table 1). Differences between 5-FU-resistant and parental cell lines were considered statistically significant at p < 0.05.

Data meeting normality and homogeneity of variance criteria were compared by Student's t-test, whereas non-normally distributed data or those with unequal variances were analyzed using the Mann–Whitney (M–W) U test. For nonparametric comparisons, box-and-whisker plots display the median and interquartile range (IQR); for parametric analyses, results are presented as mean with 95% confidence intervals (95% CIs). Three-group comparisons – examining the correlation between mRNA levels of 6 upregulated hub genes and 5-FU response rates in CCA cell lines – were performed by the Kruskal–Wallis (K–W) test followed by Dunn's multiple comparisons. The Wilcoxon matched-pairs signed-rank test was used to compare 5-FU sensitivity between resistant and parental cells. A two-sided p-value < 0.05 was considered statistically significant.

Results

Identification of 417 DEGs common to 4 stable 5-FU-resistant cancer cell lines

We searched the GEO database for transcriptome datasets of stable 5-FU-resistant cancer cell lines. Our search revealed 3 transcriptomics datasets from four 5-FU-resistant colorectal cancer cell lines and their corresponding parental cells, including GSE196900, GSE23776, and GSE81005. The DEGs analysis revealed a total of 6,965, 11,899, 2,537 and 7,255 DEGs were identified from GSE196900H (HCT116 5-FU-resistant/ HCT116 parental), GSE196900SW (SW480 5-FU-resistant/ SW480 parental cell lines), GSE23776 (MIP 5-FU-resistant/ MIP parental cell lines), and GSE81005 (HCT8 5-FU-resistant/HCT8 parental cell lines), respectively.

Volcano plots (Fig. 1) depict the distribution of significantly up- and downregulated genes in each 5-FU-resistant colorectal cancer model. By intersecting the DEG lists from all 4 datasets with a Venn diagram, we identified 417 genes that were consistently dysregulated across these resistant cell lines (Fig. 2). The complete breakdown of upregulated vs downregulated DEGs for each dataset is presented in Table 2.

Gene Ontology and pathway enrichment analysis showed several biological processes related to 5-FU resistance

Gene Ontology analysis of the 417 DEGs common to all 4 5-FU-resistant cell lines revealed significant enrichment in biological processes such as negative regulation of transcription from the RNA polymerase II promoter

(Fig. 3A). At the cellular-component level, these genes localized predominantly to the cytosol and nucleoplasm (Fig. 3B). Molecular-function analysis showed a strong bias toward protein and RNA binding activities (Fig. 3C). Reactome pathway enrichment further indicated that these DEGs participate broadly in the cellular response to stress and external stimuli, glycosylation processes, and protein translation (Fig. 3D).

Identification of the top 10 hub genes associated with 5-FU resistance using PPIs and hub gene analysis

The PPIs of 417 common DEGs associated with 5-FU resistance were constructed using STRING. From the PPIs network, the hub genes were identified and ranked as the top 30 hub genes according to their connectivity (binding scores) with other genes (Table 3). The top 10 hub genes included *HSPA8*, *UBC*, *HSPA5*, *UBB*, *RPS6*, *CYCS*, *VCP*, *EIF2S1*, *CCND1*, *RPS5*. Node colors reflect the connectivity degree, with a pseudocolor scale ranging from red to yellow, representing gene rankings from 1 to 30 (Fig. 4).

Heatmap analysis (Fig. 5) of the top 30 hub genes identified 8 candidates that were upregulated in at least 3 of the 4 5-FU-resistant cell lines: *TCP1*, *RPS6*, *RPS29*, *HSPA5*, *RPS15A*, *NOTCH1*, *CALR*, and *ACO2*. These genes were therefore designated as the commonly upregulated hubs for further investigation.

To validate the relevance of the 8 upregulated hub genes in cholangiocarcinoma, we examined their expression in Thai CCA patient tissues using GEO dataset GSE7631.[17] Six genes – *TCP1* (M–W U = –7.335, p < 0.001), *RPS6* (U = –8.521, p < 0.001), *RPS29* (U = –7.527, p < 0.001), *RPS15A* (U = –9.657, p < 0.001), *NOTCH1* (U = –3.254, p = 0.001), and *HSPA5* (t(181) = 3.505, p = 0.001) – were significantly overexpressed in tumor vs adjacent normal tissues (Fig. 6). Detailed summary statistics (median ±IQR for nonparametric tests; mean ±95% CI for parametric tests) are provided in Supplementary Table 2.

Expression of 6 upregulated hub genes associated with 5-FU resistance varied with the 5-FU sensitivity in CCA cell lines

We next evaluated whether mRNA expression of the 6 upregulated hub genes correlated with 5-FU sensitivity in 3 CCA cell lines. KKU-055, KKU-213A, and KKU-213B were treated with increasing concentrations of 5-FU for 72 h, and cell viability was assessed by MTT assay. Dose–response curves demonstrated differential inhibition of proliferation: KKU-055 exhibited a low IC_{50} and was classified as 5-FU-sensitive, whereas KKU-213A and KKU-213B displayed significantly higher IC_{50} values and were deemed 5-FU-insensitive (Fig. 7A).

Real-time PCR analysis of the 6 upregulated hub genes in CCA cell lines revealed that, of these candidates, *TCP1*
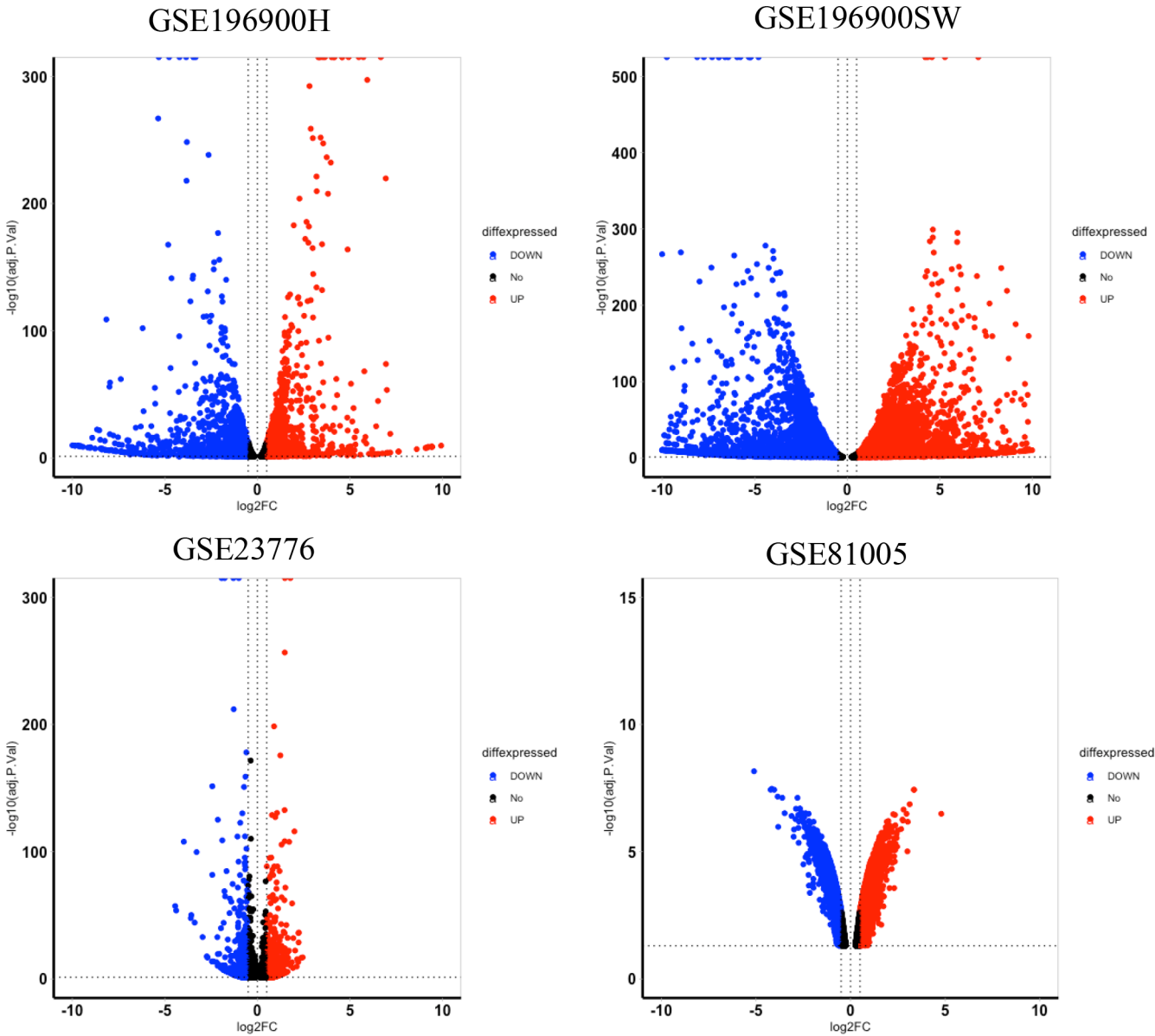
Fig. 1. Volcano plots of DEGs from 4 stable 5-FU-resistant cancer cell lines. The datasets include GSE196900H, GSE196900SW, GSE23776, and GSE81005. The X-axis represents the log2 fold change, and the Y-axis represents the negative logarithm (base 10) of the adjusted p-value. Red dots represent significantly upregulated genes, blue dots represent significantly downregulated genes, and black dots represent genes with no significant genes. GSE – GEO accession number. A p-value < 0.05 was considered statistical significance

Table 2. Transcriptomics datasets from stable 5-FU-resistant cancer cell lines

| GSE	Cell lines	Resistance ratio	Total DEGs	Up-DEGs	Down-DEGs	Experiment type
GSE196900	HCT116	49 folds	6,965	657	6,308	RNA sequencing
	SW480	35 folds	11,899	3,578	8,321	
GSE23776	MIP5	10 folds	2,537	197	2,340	RNA sequencing
GSE81005	HCT8	26 folds	7,255	688	6,567	Microarray

GSE – GEO accession number; 5-FU – 5-fluorouracil; DEGs – differentially expressed genes.

(Kruskal–Wallis H = 19.65, p < 0.001), *RPS6* (H = 19.65, p < 0.001), and *NOTCH1* (H = 17.92, p < 0.001) were significantly more abundant in the 5-FU-insensitive lines (KKU-213A and KKU-213B) than in the 5-FU-sensitive line (KKU-055) (Fig. 7B). Median expression values with interquartile ranges for all 6 genes are provided in Supplementary Table 4.

To further validate the link between mRNA expression of the 6 hub genes and 5-FU resistance, we compared parental CCA cell lines (KKU-213A, KKU-213B) with their
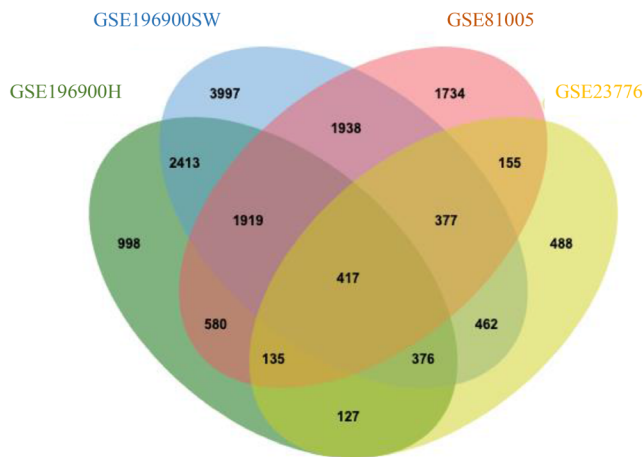
Fig. 2. Common DEGs in 5-FU-resistant cancer cell lines across 3 Gene Expression Omnibus (GEO) datasets. The Venn diagram displays the overlap of DEGs among the 3 datasets. Numbers within the sections indicate the count of DEGs specific to 1 dataset or shared among multiple datasets. GSE196900H, GSE196900SW, GSE81005, and GSE23776 are represented in green, blue, pink and yellow, respectively. The overlap number of all circles demontrates the common DEGs between the datasets

5-FU-resistant counterparts (KKU-213A-FR, KKU-213B-FR). As anticipated, the resistant sublines exhibited significantly reduced sensitivity to 5-FU vs their parental lines, as shown by Wilcoxon matched-pairs signed-rank tests (KKU-213A vs KKU-213A-FR: Z = −2.366, p = 0.018; KKU-213B vs KKU-213B-FR: Z = −2.201, p = 0.028) (Fig. 8).

Consistent with our earlier findings, *TCP1* and *NOTCH1* mRNA levels were markedly higher in the 5-FU-resistant sublines than in their parental counterparts. Specifically, *TCP1* expression was significantly elevated in KKU-213A-FR vs KKU-213A (M–W U: Z = −3.576, p < 0.001) and in KKU-213B-FR vs KKU-213B (Z = −3.582, p < 0.001). Similarly, *NOTCH1* levels were substantially increased in KKU-213A-FR compared to KKU-213A (Z = −3.580, p < 0.001) and in KKU-213B-FR compared to KKU-213B (Z = −3.576, p < 0.001) (Fig. 9). Median values with inter-quartile ranges (IQR) are detailed in Supplementary Table 5.

Potential therapeutics for *TCP1* and *NOTCH1* were identified using PanDrugs analysis; however, only *NOTCH1* has clinically approved, direct inhibitors available (Table 4).
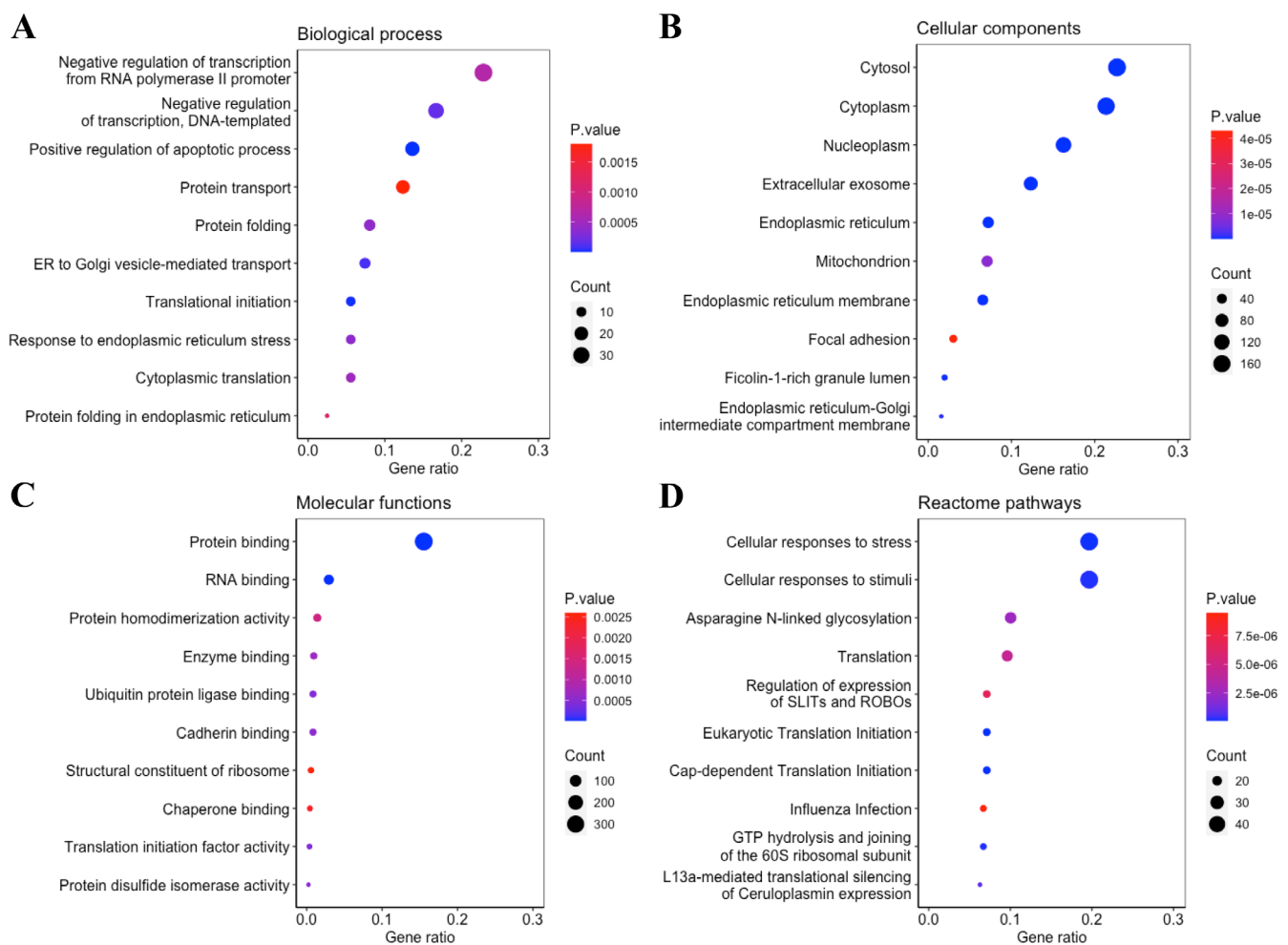
Fig. 3. Gene Ontology and pathway enrichment analysis in the common DEGs associated with 5-FU resistance. The top 10 enriched in (A) biological processes, (B) cellular components, (C) molecular function, and (D) the Reactome pathway, respectively. The size of the dots represents the count of genes, and the color indicates the p-value for the enrichment analysis

Table 3. The top 30 hub genes of common DEGs associated with 5-FU resistance

| Rank	Name	Binding scores	Rank	Name	Binding scores
1	HSPA8	55.0	16	RPL10	32.0
2	UBC	52.0	17	TCP1	32.0
3	HSPA5	52.0	18	SQSTM1	30.0
4	UBB	47.0	19	RPL38	29.0
5	RPS6	47.0	20	RPL28	29.0
6	CYCS	42.0	21	EIF2S3	27.0
7	VCP	40.0	22	RPS29	27.0
8	EIF2S1	39.0	23	CDK4	27.0
9	CCND1	39.0	24	ACO2	25.0
10	RPS5	38.0	25	PSMC2	24.0
11	EEF1A1	37.0	26	MAP1LC3B	24.0
12	RPL14	35.0	27	CALR	24.0
13	RPL15	33.0	28	EIF5B	24.0
14	RPS15A	33.0	29	NOTCH1	21.0
15	PARP1	32.0	30	ICAM1	21.0

5-FU – 5-fluorouracil; DEGs – differentially expressed genes.

Table 4. Targeted inhibitors of hub gene interactions by PanDrugs analysis

Gene(s)	Drug name	Status description	Best interaction	DScore	GScore
NOTCH1	KK8645V7LE	clinical trials	direct-target	0.514	0.7973
NOTCH1	NIROGACESTAT	clinical trials	direct-target	0.513	0.7973
NOTCH1	BRONTICTUZUMAB	clinical trials	direct-target	0.512	0.7973
NOTCH1	CRENIGACESTAT	clinical trials	direct-target	0.512	0.7973
NOTCH1	OMP-52M51	clinical trials	direct-target	0.511	0.7973

DScore – the suitability of the treatment for a particular patient; GScore – the biological relevance of a gene in the tumoral process and its druggability.
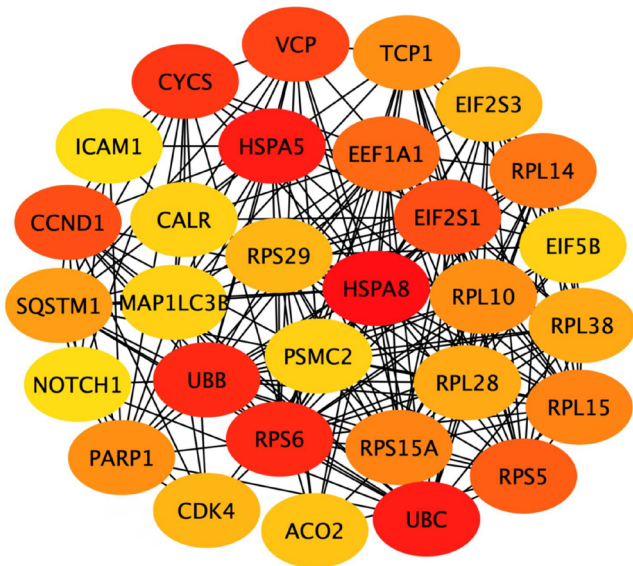
Fig. 4. Hub gene analysis of common DEGs associated with 5-FU resistance. The subnetwork shows the top 30 hub genes from the protein–protein interaction (PPI) network, generated using Cytoscape software. Node color reflects the degree of connectivity, with a pseudocolor scale from red to yellow representing the hub gene ranks from 1 to 30. Dark red color represents the highest degree while an orange color stands for the intermediate degree and yellow color is the lowest degree

Discussion

Chemoresistance to 5-FU poses a major obstacle in cholangiocarcinoma therapy.[18,19] Targeting the molecular drivers of this resistance offers a promising strategy to restore drug sensitivity. In the present study, we integrated bioinformatics analyses with PCR validation to identify TCP1 and NOTCH1 as novel mediators of 5-FU resistance in CCA. Both genes were consistently upregulated in stable 5-FU-resistant sublines compared to their parental counterparts, confirming their potential as therapeutic targets.

Chaperonin containing T-complex polypeptide 1 subunit 1 (TCP1) is a key molecular chaperone that facilitates the proper folding of nascent and stress-denatured proteins. In our study, TCP1 was markedly upregulated in 5-FU-resistant CCA cell lines, implicating it as a driver of chemoresistance. This finding is in line with reports from other malignancies, where TCP1 stabilizes oncogenic client proteins and augments pro-survival signaling. For example, in ovarian cancer, TCP1 promotes tumor cell proliferation, invasion, and migration through activation of the PI3K/AKT/mTOR pathway.[20]
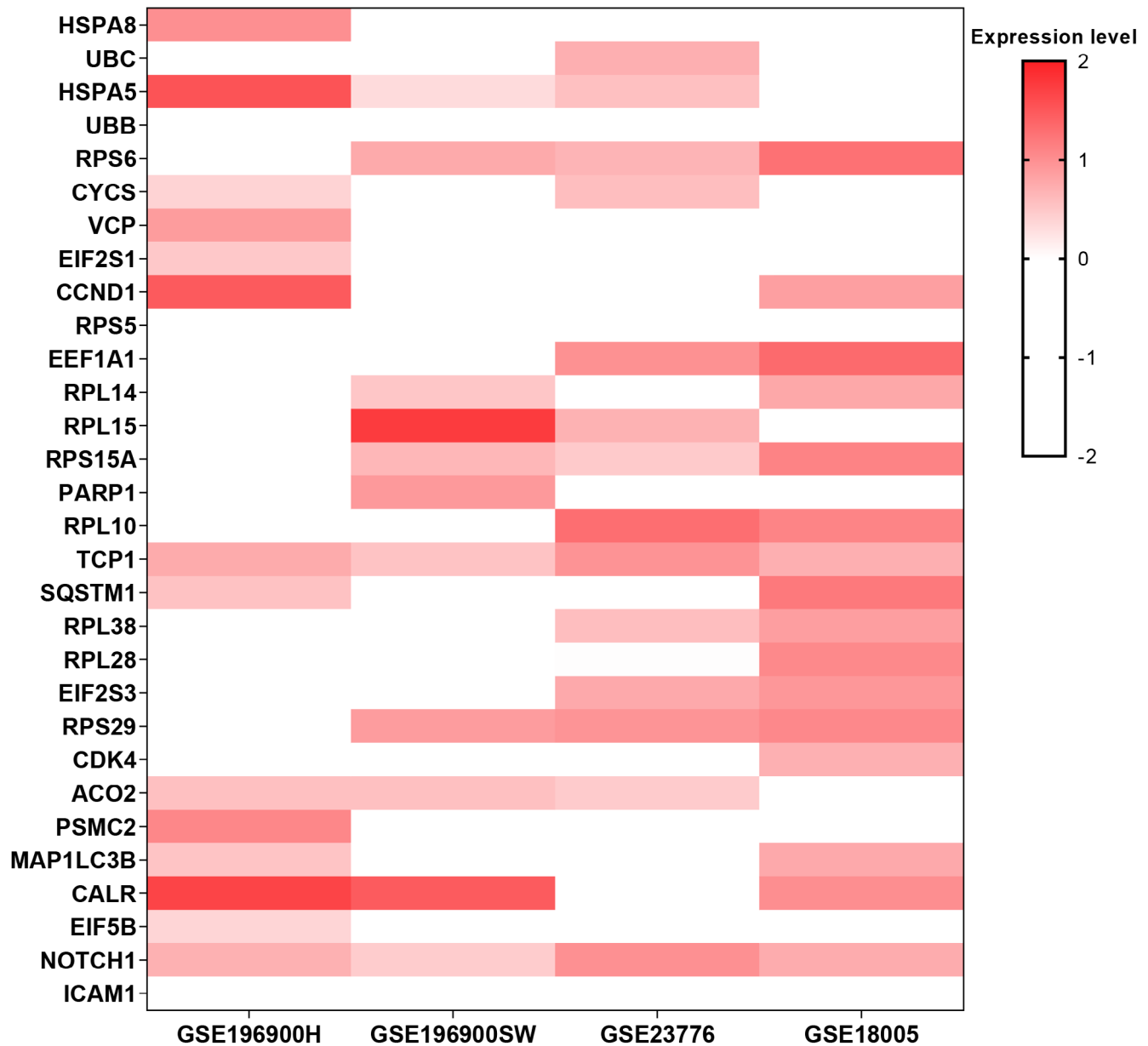
Fig. 5. Heatmap analysis of the commonly upregulated hub genes from the top 30 identified hub genes associated with 5-FU resistance. The heatmap displays the expression levels of hub genes across 4 transcriptomics datasets (GSE196900H, GSE196900SW, GSE23776, and GSE81005). The selection criteria for hub genes were those that showed upregulation in at least 3 out of the 4 5-FU-resistant cell lines. The color intensity represents the gene expression levels. Red indicates upregulation, whereas white represents downregulation

The expression of *TCP1* has been linked to upregulation of oncogenes such as *MYC*, *CCND1*, and *CDK2*, promoting breast cancer progression.[21] Although its role in 5-FU resistance remains unproven, *TCP1* overexpression correlates with poor responses to chemotherapy across multiple malignancies. Elevated *TCP1* levels confer resistance to doxorubicin and paclitaxel in breast and lung _cancer cell lines, while *TCP1* knockdown reduces X-linked inhibitor-of-apoptosis protein (*XIAP*) and β-catenin expression and inhibits metastatic behavior both in vitro and in vivo.[22]

TCP1 has been shown to enhance adriamycin resistance in acute myeloid leukemia by promoting autophagy via AKT/mTOR pathway activation. In the context of 5-FU resistance, key mediators include XIAP, β-catenin, and

the AKT/mTOR signaling axis. XIAP, a potent inhibitor of apoptosis, blocks caspase activation, and its overexpression correlates with increased resistance to both radiotherapy and chemotherapy.[23] Activation of the Wnt/β-catenin pathway has been implicated in 5-FU resistance in oral squamous cell carcinoma and colorectal cancer[24,25] while hyperactivation of AKT/mTOR signaling is a hallmark of 5-FU-resistant colorectal tumors. By targeting *TCP1*, which sits upstream of these critical pro-survival networks, it may be possible to disrupt autophagy-mediated drug resistance mechanisms and improve therapeutic outcomes in CCA and other malignancies.

Notch signaling is an evolutionarily conserved pathway that governs development, cell-fate determination,
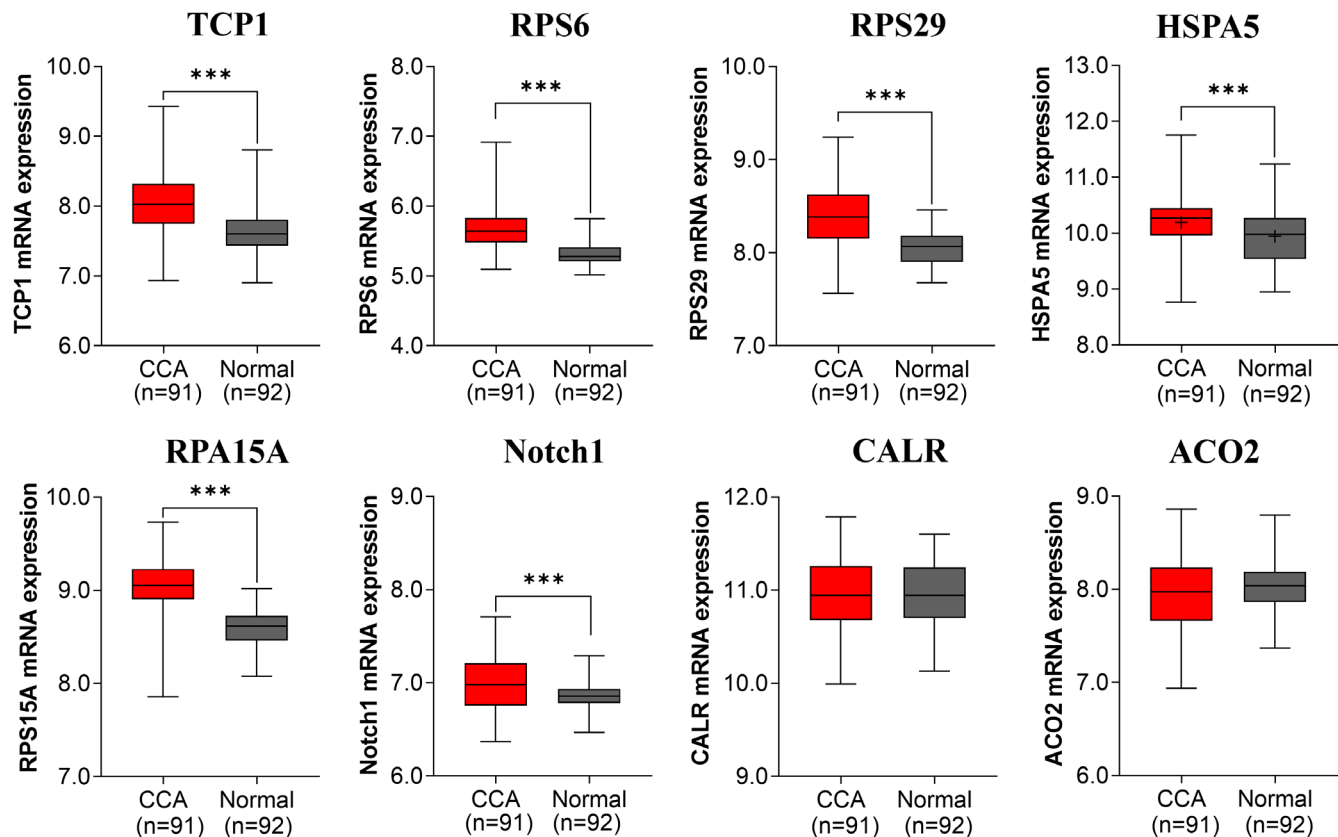
Fig. 6. The relative expression of common upregulated hub genes associated with 5-FU resistance in CCA tissues from Thai patients. Box plots show the mRNA expression levels of *TCP1*, *RPS6*, *RPS29*, *HSPA5*, *RPS15A*, *NOTCH1*, *CALR*, and *ACO2* in CCA tissues (n = 91) compared to normal tissues (n = 92). For genes analyzed using Mann–Whitney U tests, the box plots display median with interquartile range (IQR; Q1 to Q3), while for *HSPA5* (analyzed using a t-test), the box plot presents mean with 95% confidence intervals (CI). Outliers were plotted as individual points; *$p < 0.05$ was considered statistically significant (**$p < 0.01$, *** $p < 0.001$). *TCP1* – chaperonin containing t-complex polypeptide 1 subunit 1; *RPS6* – ribosomal protein S6; *RPS29* – ribosomal protein S29; *HSPA5* – heat shock protein family A (Hsp70) member 5; *RPS15A* – ribosomal protein S15A; *NOTCH1* – notch receptor 1; *CALR* – calreticulin; *ACO2* – aconitase 2

and tissue homeostasis through ligand–receptor interactions between the 4 Notch receptors (*NOTCH1–4*) and their cognate ligands (Jagged1, Jagged2, and Delta-like ligands).[26,27] Aberrant Notch activation contributes to tumorigenesis and cancer progression by upregulating oncogenic transcription factors (e.g., *MYC*, NF-κB), dysregulating cell-cycle regulators (such as p21, p27, cyclin D1, and *CCND3*), and enhancing expression of antiapoptotic proteins (including *BCL-2* and survivin).[27]

In intrahepatic cholangiocarcinoma (ICC), *NOTCH1* is markedly overexpressed compared with normal biliary epithelium, and its suppression induces apoptosis in ICC cell lines, indicating that *NOTCH1* promotes tumor cell survival by inhibiting apoptotic pathways[28]. Likewise, *JAGGED1* expression is elevated in ICC tissues relative to adjacent non-neoplastic mucosa.[29] More broadly, aberrant Notch signaling, especially via *NOTCH1*, drives chemoresistance across multiple malignancies. For example, in esophageal squamous cell carcinoma, *NOTCH1*-high KYSE70 cells display significantly greater resistance to 5-FU than *NOTCH1*-negative KYSE450 cells, and *NOTCH1* knockdown restores 5-FU sensitivity in KYSE70 cells.[30]

Knockdown of *NOTCH1* in CCA cell lines (RBE and HCCC-9810) significantly enhanced 5-FU sensitivity by downregulating the drug-efflux transporters *ABCB1* and *MRP1*, which are key mediators of chemoresistance. Given NOTCH1's role in promoting resistance to multiple chemotherapeutics, including 5-FU, its inhibition may represent a viable strategy to resensitize CCA cells and improve treatment outcomes. Several *NOTCH1* inhibitors are currently undergoing clinical evaluation, such as the γ-secretase inhibitor nirogacestat,[31] the anti-NOTCH1 monoclonal antibody brontictuzumab,[32] and the oral inhibitor crenigacestat.[33]

In our study, *NOTCH1* was markedly upregulated in 5-FU-resistant CCA cell lines compared with their parental counterparts, implicating *NOTCH1* activation in the maintenance of chemoresistance. Importantly, co-treatment with 5-FU and pharmacologic *NOTCH1* inhibitors, such as the γ-secretase inhibitor nirogacestat,[31] the anti-*NOTCH1* antibody brontictuzumab,[32] or the small-molecule crenigacestat,[33] has been shown to restore drug sensitivity in CCA models. These findings support the therapeutic potential of combining *NOTCH1* blockade with 5-FU to overcome resistance and improve clinical outcomes.

The identification of *TCP1* and *NOTCH1* as key mediators of 5-FU resistance in CCA cells opens up new avenues for therapeutic intervention. Targeting *TCP1* with
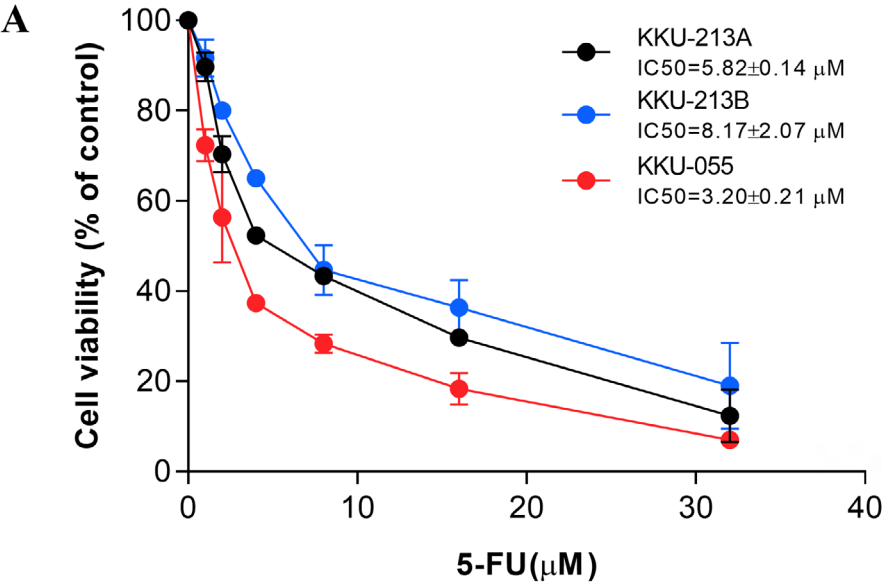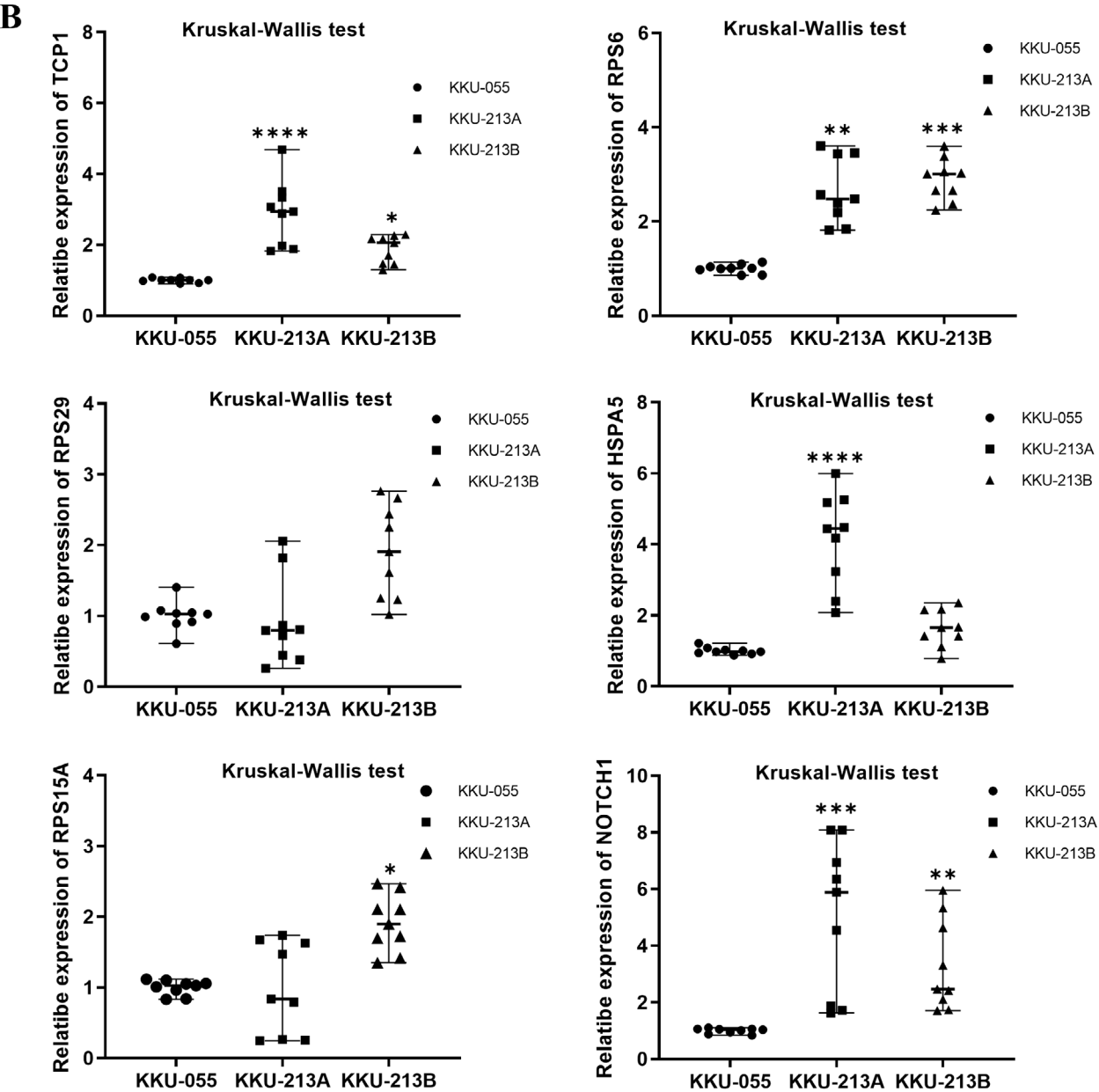
Fig. 7. The correlation between the mRNA expression levels of six upregulated hub genes associated with 5-FU resistance and the response rates to 5-FU in CCA cell lines. A. 5-FU sensitivity in KKU-055, KKU-213A, and KKU-213B represented in red, blue, and black lines, respectively; B. The expression levels of six upregulated hub genes (*TCP1*, *RPS6*, *RPS29*, *HSPA5*, *RPS15A*, and *NOTCH1*) in CCA cell lines with varying sensitivity to 5-FU. The data represent 3 independent experiments. The Kruskal–Wallis followed by Dunn's multiple comparisons test was used for statistical analysis. *p < 0.05 was considered statistically significant (**p < 0.01, ***p< 0.001, ****p < 0.0001)
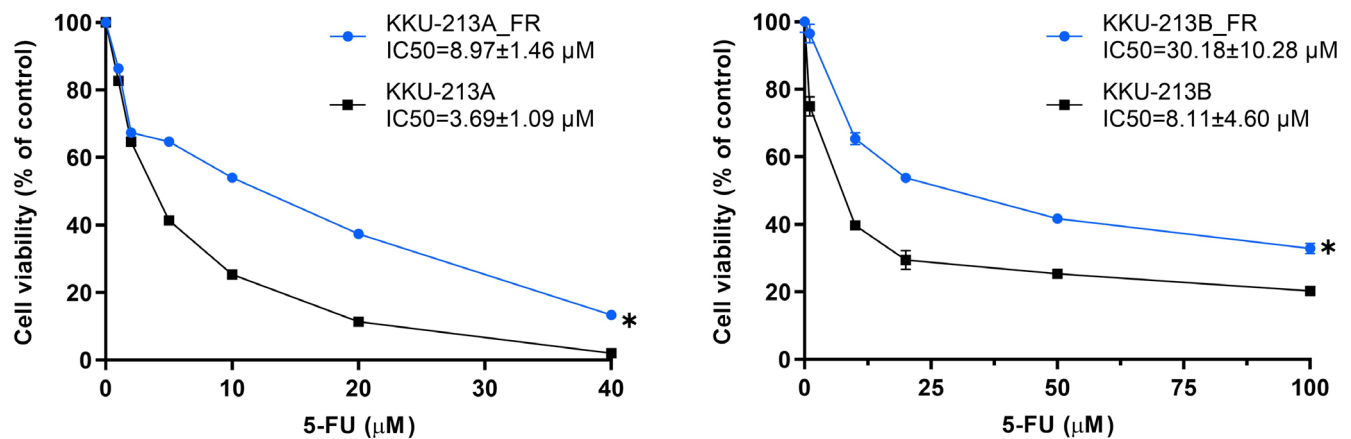
Fig. 8. 5-FU sensitivity in parental and 5-FU-resistant CCA cell lines. (Left) KKU-213A and KKU-213A_FR, and (Right) KKU-213B and KKU-213B_FR. Cell viability was measured using MTT assay after treatment with various concentrations of 5-FU for 72 h. The IC50 values represent the concentration of 5-FU required to inhibit 50% of cell growth. The data represent 3 independent experiments. Two groups were compared using the Wilcoxon matched-pairs signed rank test; *p < 0.05 is considered as the statistical significance
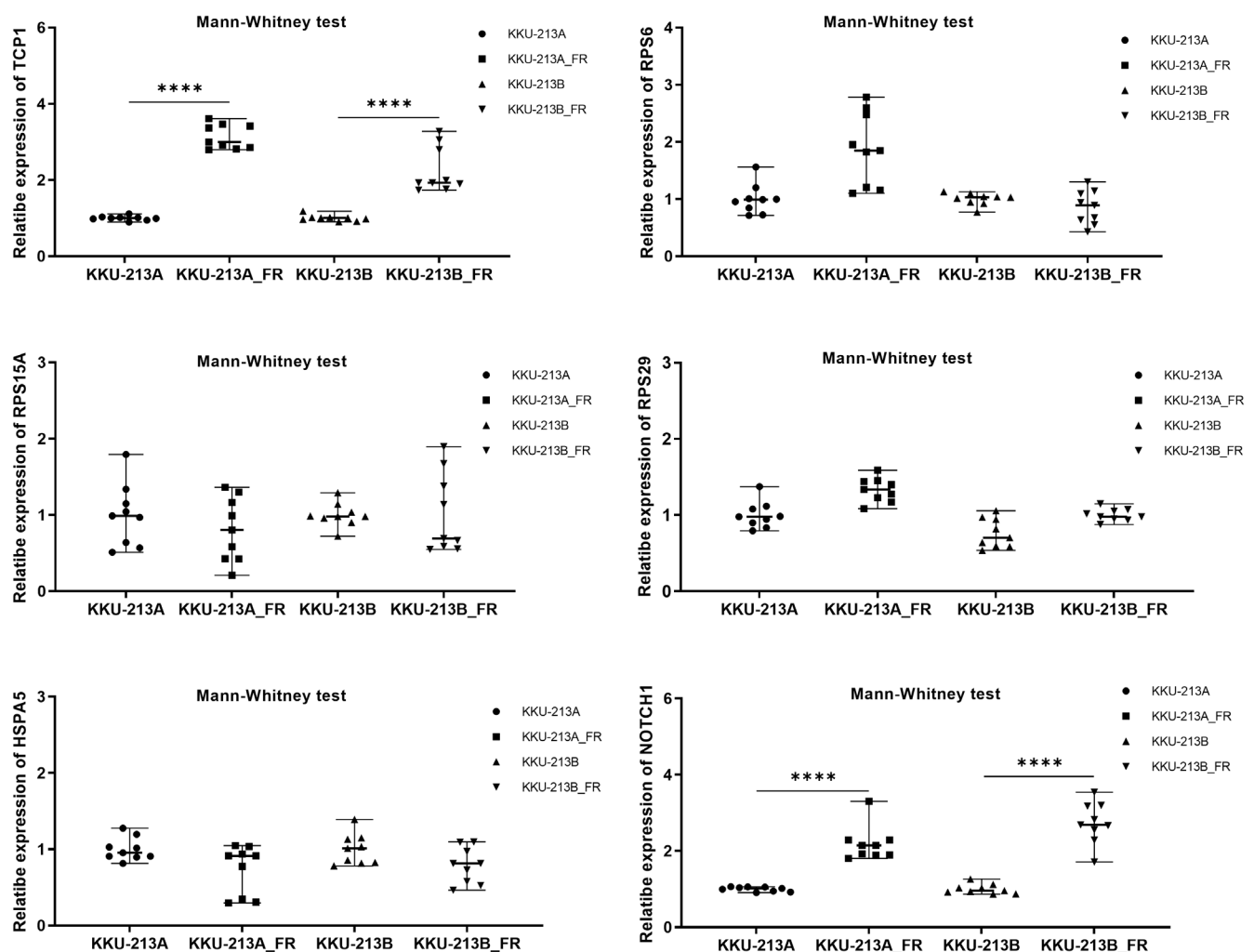

Fig. 9. The correlation between the mRNA expression levels of 6 upregulated hub genes associated with 5-FU resistance in parental CCA cell lines (KKU-213A and KKU-213B) compared to their corresponding stable 5-FU-resistant cell lines (KKU-213A-FR and KKU-213B-FR). The results show significantly higher mRNA levels of TCP1 and NOTCH1 in the stable 5-FU-resistant CCA cell lines compared to the parental cell lines. Mann–Whitney U test was used to evaluate the expression differences between 5-FU-resistant and parental; *p < 0.05 is considered as the statistical significance (**p < 0.01, ***p< 0.001, ****p < 0.0001)

small-molecule inhibitors or RNA-based therapies could disrupt its role in stabilizing oncogenic proteins and sensitizing cancer cells to 5-FU. Similarly, the use of *NOTCH1* inhibitors, such as brontictuzumab or crenigacestat, which are currently in clinical trials for other cancers, may enhance the efficacy of 5-FU in CCA. Combination therapies that include 5-FU and inhibitors of *TCP1* or *NOTCH1* could be explored in preclinical models to evaluate their potential to overcome chemoresistance and improve treatment outcomes.

Limitations

Further research is required to validate these findings in the tissues of CCA patients with resistance to 5-FU treatment vs 5-FU sensitive patients.

Conclusions

This study is the 1st report to identify *TCP1* and *NOTCH1* as key molecules associated with 5-FU resistance in CCA. Our findings suggest that the overexpression of these genes contributes to chemoresistance through mechanisms such as protein folding, cellular stress response, and drug efflux regulation. Importantly, targeting TCP1 and *NOTCH1* holds promise as a strategy to overcome 5-FU resistance and improve therapeutic outcomes in CCA patients. Future studies should validate these findings in clinical samples and explore the efficacy of combining *NOTCH1* inhibitors with 5-FU in preclinical models. These efforts could pave the way for novel therapeutic interventions that enhance the effectiveness of chemotherapy and address the pressing challenge of chemoresistance in CCA.

Data availability statement

The transcriptomic datasets from stable 5-FU-resistant cancer cell lines and Thai CCA patient tissues were retrieved from the GEO database and are openly available in Figshare [https://dx.doi.org/10.6084/m9.figshare.27139791].

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