

Integrative analyses of noncoding RNAs reveal the potential mechanisms augmenting tumor malignancy in lung adenocarcinoma

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ABSTRACT

Precise noncoding RNA (ncRNA)-based network prediction is necessary to reveal ncRNA functions and pathological mechanisms. Here, we established a systemic pipeline to identify prognostic ncRNAs, predict their functions and explore their pathological mechanisms in lung adenocarcinoma (LUAD). After *in silico* and experimental validation based on evaluations of prognostic value in multiple LUAD cohorts, we selected the *PTTG3P* pseudogene from among other prognostic ncRNAs (*MIR497HG*, *HSP078*, *TBX5-AS1*, *LOC100506990* and *C14orf64*) for mechanistic studies. *PTTG3P* upregulation in LUAD cells shortens the metaphase to anaphase transition in mitosis, increases cell viability after cisplatin or paclitaxel treatment, facilitates tumor growth that leads to poor survival in orthotopic lung models, and is associated with a poor survival rate in LUAD patients in the TCGA cohort who received chemotherapy. Mechanistically, *PTTG3P* acts as an ncRNA that interacts with the transcription factor FOXM1 to regulate the transcriptional activation of the mitotic checkpoint kinase BUB1B, which augments tumor growth and chemoresistance and leads to poor outcomes for LUAD patients. Overall, we established a systematic strategy to uncover prog-

nostic ncRNAs with functional prediction methods suitable for pan-cancer studies. Moreover, we revealed that *PTTG3P*, due to its upregulation of the *PTTG3P/FOXM1/BUB1B* axis, could be a therapeutic target for LUAD patients.

INTRODUCTION

According to the annotations generated by the GENCODE and ENCODE consortia, 75% of the human genome has been estimated to be transcribed into RNAs in different cell types and tissues (1,2). Approximately one-third of these complex transcripts are categorized as coding RNAs with various splicing isoforms, and the remaining two-thirds are categorized as noncoding RNAs (ncRNAs), which comprise long noncoding RNAs (lncRNAs), pseudogenes and other structural and regulatory small RNAs (3). Emerging evidence suggests that ncRNAs play critical roles in a broad range of biological functions to regulate transcription, gene expression and protein signaling via participation in histone modification and transcriptional protein complexes, the remodeling of 3D nuclear organelles, the efficacy of protein translation via direct binding to mRNAs, and the modulation of cellular signaling via scaffolding protein complexes (4,5). In comparison to coding genes that are well annotated with the biological functions and pathological connections to human diseases and symptoms, the molecular functions and pathological mechanisms of abundant expressed ncRNAs remain elusive. Since alterations of some ncRNAs

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were found to be associated with disease progression and patient outcomes, especially in cancers, aberrant ncRNAs might serve as biomarkers and therapeutic targets for potential clinical interventions (6). Nevertheless, the lack of systematic studies that predict the biological functions and prognostic values of ncRNAs has hampered progress in the development of mechanistic understanding and clinical interventions utilizing ncRNAs to prolong the lives of cancer patients.

Lung cancer is the most common cancer in humans, and there were an estimated 2 million cases and 1.7 million deaths due to lung cancer worldwide in 2018 (7). Lung adenocarcinoma (LUAD) is the predominant subtype and accounts for 40% of lung cancers for which there are abundant resources including genomic information accompanied with patient clinical features in the public domains for integrated genomic analysis (8). With the conventional focus on mutated coding genes in LUAD for target therapies, some bioinformatic efforts were conducted to identify prognostic coding and non-coding gene signatures but lack of mechanistic and experimental validations for potential translational applications (9–11).

To comprehensively explore prognosis-associated ncRNAs by utilizing functional prediction and mechanistic validations, we established a systematic pipeline to identify prognostic ncRNAs and their functions and mechanisms in LUAD. We hypothesized that the co-expression of coding genes with ncRNAs could precisely predict the unknown functions of the ncRNAs based on the presence of the same biological condition in a given dataset. After determining the association of aberrantly expressed genes with LUAD patient prognosis, the networks containing prognostic ncRNAs with highly correlated coding genes that existed in at least half of the LUAD datasets were selected for functional and mechanistic predictions. We identified six prognostic ncRNA-associated networks and selected the most consistent prognostic ncRNA, *PTTG3P* (pituitary tumor-transforming gene 3 pseudogene), for experimental and mechanistic validations of its correlation with poor prognosis in LUAD patients.

PTTG3P is located in chromosomal region 8q13 and has been annotated as a processed, intronless pseudogene with very high RNA sequence similarity to its ancestral genes *PTTG1* and *PTTG2*, which are located in chromosomal regions 5q33 and 4p12, respectively (12,13). *PTTG1* is an aberrantly expressed oncogenic securin protein that has a negative regulatory effect on p53 in modulating chromosome stability and DNA repair in cancers (14–16). In contrast, although a few studies have noted differences in the expression and function of *PTTG* family members in different cancers (17–19), neither *PTTG2* nor *PTTG3P* has been shown to participate in tumor progression in LUAD. In our study, consistent with the functional and pathological predictions that *PTTG3P* participates in mitosis and drug resistance, *PTTG3P* ncRNA forms a complex with the *FOXM1* transcription factor to target the promoter of *BUB1B* (mitotic checkpoint serine/threonine kinase B). The activation of the *PTTG3P* ncRNA/*FOXM1*/*BUB1B* axis shortens the metaphase-anaphase transition, which increases cell proliferation, tumor growth and drug resistance and leads to poor survival in LUAD patients. Overall, we

established a comprehensive and systematic pipeline that is applicable to other cancer types to identify prognostic driver ncRNAs, to predict their biological functions, and to validate and reveal the molecular pathological mechanisms that underlie the involvement of ncRNAs in tumor progression, which could aid in the development of potential clinical interventions.

MATERIALS AND METHODS

Dataset collection

The cohort datasets of transcriptomes of LUAD patients were downloaded from Genome Expression Omnibus (GEO). The identification of differentially expressed probes (DEPs) was conducted by using four datasets (GSE27262, GSE31210, GSE30219 and GSE19188) containing both the tumor and normal tissue samples. Six datasets (GSE31210, GSE50081, GSE37745, GSE30219, GSE3141 and GSE19188) with 10-year survival information of LUAD patients were selected to conduct the survival analysis. The microarray datasets were all normalized and downloaded from GEO, which were generated using Affymetrix Human Genome U133 Plus 2.0 Arrays containing 39455 coding and 2038 non-coding probes, and summarized in Supplementary Table S1. The level 3 gene expression data of LUAD RNA-Seq was downloaded from The Cancer Genome Atlas (TCGA).

Identification of prognostic differential expressed probes in LUAD

The DEPs in tumor samples were selected independently in each of four datasets via *F*-test and *t*-test with FDR (false discovery rate) <0.05 and cross-validated in between datasets deposited from different laboratories for reducing the false positive possibility. For prognostic DEPs, we conducted the cox-regression test on six datasets with 10-year survival information by using the function *coxph* in survival R package (20). We identified 627 prognostic DEPs with criterion of each possible candidate in the log-rank test *P*-value < 0.05 by computing univariate cox analyses and were presented in at least three different datasets.

Identification of ncRNA-associated networks with a functional prediction method

Modules were constructed from co-expression networks with the weighted gene co-expression network analysis (WGCNA) package by using the standard procedure (21). A strong correlation was obtained by weighting the correlation based on power β derived from the power adjacency function. The approximate scale-free topological overlap of each dataset was represented by an adjacency matrix by using a well-defined formula for weighted networks and applied to average linkage hierarchical clustering through topological overlap dissimilarity (1-topological overlap). According to the hierarchical clustering, the probes were grouped as a module with a height cut-off for the dynamic tree of 0.95 and the minimum number of probes in a module was 5.

For precise functional prediction, the significance of a gene identified by the WGCNA was calculated by the $-\log(P\text{-value of the log-rank test})$ and the k_{Within} value to demonstrate the connectivity of the probes within the modules. In this study, the top 15 correlated DEPs in each dataset were ranked on the basis of the k_{Within} values to identify the downstream targets of the specific ncRNAs. After extracting each prognostic ncRNA with its co-expressed coding genes from specific modules in at least three datasets, the prognostic ncRNA networks were applied to pathway enrichment analysis by WebGestalt (a web-based gene set analysis toolkit).

LUAD Tissue arrays and RNA-ISH

Tumors tissues of Taiwan LUAD patients were collected and spotted onto microarray by Dr Teh-Ying Chou from Taipei Veterans General Hospital (VGH) (22). LUAD patients enrolled from 2002 to 2006 at Taipei VGH in the surgical pathology archives were included in this study. Preparation of tissue arrays and patient demography were provided in the Supplementary Methods and Supplementary Table S2, respectively.

Probes for *PTTG3P* RNA were designed through LNATM Probe Designer (<https://www.exiqon.com/mRNA-probes>), and the sequences of probes and the competition RNAs for assays are provided in the Supplementary Table S3. After dewaxing, the tissue arrays were treated with protease K at 37°C for 10 min, fixed by 4% paraformaldehyde at room temperature for 15 min and then hybridized by LNA probes at 60°C overnight. Arrays were washed with 0.2× SSC buffer with 2% BSA at 4°C for 10 min to remove non-specific probes by following the manufactural protocol of IsHyb In Situ Hybridization (ISH) Kit (BC-K2191050, Blossom Biotechnologies). After arrays dehydration and sealing, expression intensity of *PTTG3P* was determined by the staining intensity with *H*-score (intensity scores multiplied by the number of percentage of cells with positive *PTTG3P* staining) by two pathologists. The high and low expression of *PTTG3P* were grouped based on high and low tertiles of total *PTTG3P* *H*-score for survival analysis.

LUAD cell lines

Eight lung cell lines including immortal lung epithelial NL-20 and LUAD cell lines H1299, A549, CL1-0, CL1-5, H23, H1435 and H1437 were described previously (23,24). All cell lines were cultured in Dulbecco's modified Eagle's medium (DMEM) with 10% FBS and incubated in 5% CO₂ at 37°C.

Transfections with knockdown or overexpression reagents

Knockdown of *PTTG3P* in H1299 and CL1-5 was conducted by the CRISPR interference CRISPRi (RNAi Core of Academia Sinica in Taiwan), and the sequence of sgRNA are provided in Supplementary Table S3. At first, each sgRNA vector (*PTTG3P* sgRNA-1, *PTTG3P* sgRNA-2 and sgCtrl) were transfected with p5w-dCas9-KRAB.pBsd plasmid into H1299 and CL1-5 cell lines by

standard protocol of jetPRIME (PolyPlus). After 2 days transfection, we obtained the specific clonal population by single cell dilution in 96-well plates with selection of 2 μg/ml puromycin and 10 μg/ml Blastidin (Thermo Fisher ScientificTM) for at least 2 weeks.

H1299 cells were infected with luciferase-expressing lentivirus for an lung orthotopic model and selected in the presence of 600 μg/ml G418 (Thermo Fisher ScientificTM). Luciferase expression vector of psPAX2 and pMD2.G were transfected into 293T by jetPRIME for virus production. After 2 days, the virus containing supernatant was collected by passing through the 0.22 μm filters with polybrene ready for H1299 infection. Plasmids for overexpression were transfected into cells by following the standard protocol of jetPRIME, and the details for plasmid construction are provided in the Supplementary Methods.

Lung orthotopic mouse models

The 5-week old male NOD-SCID/CB.17 mice were purchased from LASCO, Taiwan. Cells (H1299: 1 × 10⁶/50 μl; A549: 3 × 10⁶/50 μl) mixed with Matrigel (BD) were injected into the left lung of anesthetized mice (6–8-week old) as described in the previous studies (25). Tumor tissue blocks and slides of orthotopic lung were prepared from left lung of NOD/SCID mice by the Pathology Core of IBMS (Academia Sinica) for staining with *PTTG3P* RNA-ISH probe or H&E.

Subcellular fractionation, RNA extraction and RT-qPCR

Subcellular fractionation was conducted by following the standard protocol (26). Total RNA samples were extracted by Trizol (Thermo Fisher ScientificTM) based on the recommended protocol and cDNA were generated by Maxima H Minus First Strand cDNA Synthesis Kit (Thermo Fisher ScientificTM) followed by the manufacturer's instructions. For detecting RNA expression, equal amounts of cDNA mixed with SYBR Green Master Mix (Thermo Fisher ScientificTM) and primers was performed in Applied Biosystems 7500 Real-Time PCR System. 18S ribosomal RNA (rRNA) served as an internal control for real-time qPCR analyses. See Supplementary Tables S3 for sequences of RT-qPCR primers.

Cell proliferation and cell viability assays

Cell proliferation assay was performed by xCELLigence RTCA system and followed the standard protocol of xCELLigence instrument. In general, 2500 cells were seeded in each well of a 16-well E-plate and monitored the cell index of each hour. The cell viability assay detected by Alamar blue reagent (Thermo Fisher ScientificTM) according to manufacturer's instruction were provided in the Supplementary Methods. Cisplatin is gifts kindly provided by Dr Tang Tang of our institute.

Flow cytometry

Cells (1 × 10⁶) were washed by phosphate buffered saline (PBS) twice, and then fixed with 70% ice-cold ethanol at

–20°C overnight. The fixed cells were washed by PBS and incubated in the dark place containing 400 µg/ml RNase A, 20 µg/ml propidium iodide (PI), 0.1% Triton X-100 in PBS at room temperature for at least 90 minutes. After PBS washes, cells were visualized by BD LSR II flow cytometer.

Live cell time-lapse imaging

Cancer cells were seeded into the 12-well plate with 80% confluence and replaced medium containing 65 ng/ml nocodazole in the next day. After treated with nocodazole for 18 h, cells were treated with Hoechst 33342 (Sigma-Aldrich), washed by PBS in at least three times and replaced with fresh medium to monitor live cells by Leica DMI 6000B Live-Cell Fluorescence Observation System.

Western blot

Cells were lysed with RIPA buffer containing protease inhibitor cocktail (Roche) and phosphatase inhibitor (Millipore). Equal amount of lysates were loaded onto a 10% SDS-PAGE gel and blotted onto PVDF membranes. After treated with 5% blocking buffer (non-fat dry milk dissolved in PBST with 0.1% Tween-20) for 30 min in room temperature, western blotting was incubated with primary antibodies against HA-tag (Covance) or β-actin, or GAPDH (proteintech). After 2-hour incubation with primary antibodies, membranes were washed by 0.1% PBST three times followed by incubated with peroxidase-conjugated rabbit/mouse secondary antibodies for detection. The chemoluminescence was detected by ImageQuant LAS 4000 mini.

RNA-seq

RNA samples were extracted by NucleoSpin RNA (MN) kit and the transcriptome were performed by next generation sequencing (NGS) and analyzed by Novogene (BIOTOOLS). In brief, libraries were generated by using the Agilent Sure Select Strand Specific RNA Library Prep (Illumina) kit after rRNA removed from RNA samples. NGS sequencing were performed in HiSeq4000 (Illumina) for 150 bp paired-end reads. The detailed versions of software and analysis are provided in the Supplementary Table S4. The transcriptome data was deposited into GEO database with accession number GSE114826.

Reporter constructs and luciferase assays

The BUB1B promoter fragment (–3000/+39, –585/+39, –236/+39) were PCR amplified from A549 genomic DNA incorporating NheI and XhoI cutting sites, and then PCR products were inserted into pGL3 Luciferase Reporter Vectors (Promega). The primers of promoter fragments were provided in Supplementary Table S2.

Reporter constructs containing BUB1B promoter fragment were co-transfected with pRK5 Renilla reporter vector as control of transfection efficiency. After 36–48 h, the luciferase assay was followed the standard protocol of Dual-Glo[®] Luciferase Assay System (Promega), and detected by the recommended program of (TECAN)-Infinite[®] M1000 PRO. The luciferase activity was normalized to the Renilla activities.

Chromatin-immunoprecipitation (ChIP) and RNA-immunoprecipitation (RIP)

DNA samples were prepared by following the standard protocol of EpiTect ChIP OneDay Kit (Qiagen), and the RIP assay was based on the protocol as previous study (27). The ChIP-grade primary antibodies against FOXM1 or IgG were ordered from Cell Signaling Technology. The primer sequences for cloning the promoter regions of BUB1B were provided in Supplementary Table S3.

RNA fluorescence in situ hybridization (RNA-FISH)

Cells were cultured on the silane-coating slides for 48 hours with 50% confluence. After washing cells with PBS, cells were fixed in 4% paraformaldehyde for 20 min at room temperature. After washing with DEPC-treated PBS, cells were incubated with 0.2 M HCl for 20 min at room temperature. Cells were further treated with protease K at 37°C for 10 min after washing with DEPC-treated PBS, and then further hybridized by LNA probes at 60°C overnight. Slides were washed with 0.2xSSC buffer with 2% BSA at 4°C for 10 min to remove non-specific probes, and further incubated with anti-DIG-HRP (NEF832001EA, Blossom Biotechnologies) for 1 h. After washing cells by DEPC-PBS, fluorescence were amplified by following the manufacturing protocol of TSA Plus Fluorescence Kits (PK-NEL705A, Blossom Biotechnologies). After slides were counterstained with Hoechst dye 33342 and mounted, we examined the intensity of *PTTG3P* signaling by confocal microscope LSM510.

RNA pull-down

Cells were synchronized to G2/M by 65 ng/ml nocodazole, and RNA pull-down assay was performed by following the previous study (28). *In vitro* transcription of biotinylated T7-based or SP6-based RNA were prepared by AmpliScribe[™]T7-Flash[™]Biotin-RNA Transcription Kit (Lucigen) and SP6 RNA polymerase with Biotin RNA labeling Mix (Roche).

Chromatin Isolation by RNA purification (ChIRP)

ChIRP was performed by following the previous protocol (29). In the beginning, we designed anti-sense oligo probes of *PTTG3P* and LacZ with biotin-labeled at 3'-prime end by ChIRP Probe Designer. Forty million H1299 cells were treated with 65 ng/ml nocodazole for G2/M synchronization followed by 1% glutaraldehyde for 10 minutes at room temperature for crosslinking. After lysing crosslinking, DNA was sheared to 200–500 bp at 4°C and hybridized with biotinylated target probes in 37°C. Chromatin bound with probes were isolated by Streptavidin Mag Sepharose (GE Healthcare) and individually extracted DNA, RNA, and protein for further detection as previously described (29). Sequences of biotinylated probes used in ChIRP were listed in Supplementary Table S3

Statistical analysis

Differentially expressed probes (DEPs) were selected through Student's *t*-test, *F*-test and adjust *P*-value by

Benjamini–Hochberg (BH) correction. Statistical significance of triplicates for Student's *t*-test, one-way or two-way ANOVA, and survival analysis was calculated by GraphPad Prism version 8.

RESULTS

Identification of prognostic ncRNAs in LUAD patients with predicted biological functions and drug responses

To discover prognostic ncRNAs with their biological functions and possible clinical applications in LUAD patients, we applied a systematic approach to identify and predict the functions and drug sensitivities of ncRNAs that were highly correlated with patient survival in LUAD (Figure 1A). We identified 5647 differentially expressed probes (DEPs) in the tumor and normal samples (Supplementary Table S1), including 627 DEPs that were associated with patient survival. These 627 DEPs were examined the enrichment analyses and were highly enriched in the reported gene signature of poor survival in NSCLC (Supplementary Figure S1A) (30), and then 627 prognostic DEPs, including 618 coding and 9 noncoding DEPs, were further applied to identify the prognostic gene networks by constructing co-expression networks with the weighted gene co-expression network analysis (WGCNA) (Supplementary Figure S1B–D) (see details in the Materials and Methods section) (21). We found six ncRNA-containing networks, including modules containing *PTTG3P*, *MIR497HG*, *HSP078*, *TBX5-ASI*, *C14orf64* and *LOC100506990* that were associated with patient survival. Each ncRNA and its associated module maintained their association with prognosis in at least half of the datasets, as indicated by a statistically significant *P*-value based on the log-rank test (Figure 1B; Supplementary Figure S1E; Tables S5 and S6). We noticed that two datasets, GSE3141 and GSE19188, showed no statistical association between the six ncRNAs and LUAD patient survival, and these two datasets had relatively small sample sizes compared to those of the other datasets. Our results indicated that the upregulation of *PTTG3P* and its associated network was correlated with poor prognosis in LUAD patients, while the rest of the ncRNAs tended to be associated with a good prognosis in LUAD patients.

After identifying prognostic ncRNAs and their highly correlated networks according to stringent criteria, gene enrichment analysis was applied to each ncRNA-containing network by the web-based gene set analysis toolkit (WebGestalt) to investigate their potential biological functions and involvement in drug responses (Figure 1C, D) (31). Consistently, although only in a relatively small number of prognostic probes were used, several predicted biological processes, such as the nicotine degradation II pathway, axon guidance, the Slit-Robo pathway, sema4D-related and integrin-related cell migration, and developmental biology were demonstrated to contribute to lung cancer tumor progression (Figure 1C) (32–35). Interestingly, our statistical analysis showed that the *PTTG3P*-containing network was significantly enriched in cell cycle-related pathways, especially those related to mitosis. This implied that aberrant *PTTG3P* expression might lead to a poor prognosis by disturbing the process of mitosis in LUAD. Moreover, drug association analysis showed that the *PTTG3P*-containing

network was significantly enriched in genes associated with first-line clinical chemotherapeutic drugs such as cisplatin and paclitaxel in LUAD patients (Figure 1D).

In silico and experimental validations of ncRNAs associated with poor survival in LUAD patient data retrieved from TCGA

To validate the aberrant expression and prognostic value of the six prognostic ncRNAs *in silico*, we examined the association of their expression with LUAD patient survival with RNA-Seq data from TCGA that was normalized by either Firebrowse or MiTranscriptome (36,37). Except for the lack of annotation of *LOC100506990* in both datasets, we found consistently aberrant expression and similar trends in the hazard ratios for the remaining five ncRNAs in LUAD patient data collected from TCGA (Figure 2A, B). Experimentally, we also validated the altered expression of the six ncRNAs with cDNAs generated from paired LUAD samples by RT-qPCR (Figure 2C; Supplementary Table S2). Notably, *PTTG3P* showed the most consistent aberrant expression and statistical association with poor prognosis in LUAD, which was validated in the independent and LUAD datasets in the TCGA (Figure 2C–E), suggesting that *PTTG3P* could serve as a prognostic oncogenic marker in LUAD. Consistently, increasing expression of *PTTG3P* is also predicted in association with poor patient survival of LUAD independently analyzed by The Atlas of Noncoding RNAs in Cancer (TANRIC) (38).

PTTG3P was shown to be upregulated and associated with poor survival by using RNA-ISH in tumors from LUAD patients in Taiwan

After *in silico* validation of the six prognostic ncRNAs, we selected the *PTTG3P* ncRNA for biological and pathological validations based on its consistency and the stringent criteria described above. Owing to the high sequence identity among members of the PTTG family and the clarification of the role of *PTTG3P* RNA expression in LUAD, we found that both PTTG1 and *PTTG3P* but not PTTG2 were upregulated in LUAD tissues in the TCGA dataset (Supplementary Figure S2A). The expression of upregulated *PTTG3P* showed no correlation with the different stages of LUAD in tissues (Supplementary Figure S2B). Since PTTG1 is upregulated in LUAD tissues, as shown in previous reports (15,16), we examined the aberrant expression and prognostic value of *PTTG3P* with RNA *in situ* hybridization (RNA-ISH) in LUAD patients from Taiwan with tissue arrays (Supplementary Table S2). We designed a *PTTG3P*-specific RNA-ISH probe and competing PTTG1, PTTG2, and *PTTG3P* RNAs (Figure 2F, upper panel). After competition with different concentrations of competing RNAs (Figure 2F, lower panels; Supplementary Figure S3), we demonstrated that the *PTTG3P*-specific probe could specifically recognize *PTTG3P* expression with high specificity in tissue arrays of LUAD. Based on the RNA-ISH experiments, we consistently showed that *PTTG3P* is indeed upregulated and is not correlated with LUAD stages (Supplementary Figure S2C), but it is associated with poor survival of LUAD patients in Taiwan (Figure 2G).

Inhibition of *PTTG3P* perturbed cell mitosis, proliferation and drug sensitivity in LUAD

Since *PTTG3P* could play a role in modulating the cell cycle (Figure 1C), we next examined the predicted functions of *PTTG3P* in LUAD cell lines. We investigated the RNA expression of *PTTG3P* in LUAD cell lines with RT-qPCR assays. We found that *PTTG3P* had relatively low expression levels in the immortalized lung cell line NL-20 and in the cancer cell lines A549 and H23; moderate expression levels in the H1435, H1437 and CL1-0 cell lines; and high expression levels in the CL1-5 and H1299 LUAD cell lines (Supplementary Figure S2D). Based on these results, we selected the H1299 and CL1-5 cells for *PTTG3P* knockdown experiments and the A549 and CL1-0 cell lines for *PTTG3P* overexpression experiments for functional assays.

Because of the high sequence identity of PTTG family members but their different chromosomal locations, we performed *PTTG3P*-specific knockdown via CRISPR interference (CRISPRi) (39). Two *PTTG3P*-specific single guide RNAs (sgRNAs) (sg*PTTG3P*-1 and sg*PTTG3P*-2) were designed and cotransfected with dCas9-KRAB into H1299 and CL1-5 cells to repress *PTTG3P* expression at the transcriptional level. Since PTTG2 was shown to be expressed at low levels in normal lung and cancer tissues without activation during LUAD progression in the TCGA dataset (Supplementary Figure S2A), we compared the knockdown specificity of sg*PTTG3P*-1 and sg*PTTG3P*-2 by verifying the RNA expression of PTTG1 and *PTTG3P* in LUAD cell lines. Our results showed that *PTTG3P* knockdown significantly decreased *PTTG3P* expression with no significant interference in the expression of PTTG1 in the H1299 and CL1-5 cell lines (Figure 3A).

To investigate the involvement of the aberrant upregulation of *PTTG3P*, as predicted by the mitotic cell cycle, in LUAD progression, we examined the downregulation of *PTTG3P* expression in the processes involved in mitosis, cell proliferation and drug sensitivity. We found that cell proliferation was decreased in *PTTG3P*-knockdown H1299 and CL1-5 cells compared with that in cells treated with control sgRNA (Figure 3B). When examining the effects of *PTTG3P* knockdown on the percentages of cells distributed in different phases of the cell cycle after serum-free synchronization, we found that most of the H1299 cells entered the G2/M phase after approximately 8 hours when treated with control sgRNA and entered the G2/M phase after approximately 10 hours upon *PTTG3P* knockdown after replacement with fresh medium (Figure 3C). Moreover, the population of cells in the G1 phase increased after 10 hours in cells transfected with control sgRNA but not in cells transfected with sgRNA targeting *PTTG3P*, indicating that the depletion of *PTTG3P* affected cell cycle progression.

To dissect the mechanism involved in mitotic cell cycle interference in *PTTG3P* knockdown cells, we monitored the transition from metaphase to anaphase in mitosis with Hoechst dye staining and fluorescence microscopy. Our results indicated that *PTTG3P* knockdown in H1299 and CL1-5 cells significantly lengthened the time required for the metaphase to anaphase transition by two-fold compared with that in cells treated with control sgRNA (Figure 3D, E).

To further investigate the effect of mitotic interference by upregulated *PTTG3P* on drug sensitivity, we examined the effect of *PTTG3P* knockdown on viability in H1299 cells treated with the common LUAD chemotherapeutic drugs paclitaxel and cisplatin. We found that the knockdown of *PTTG3P* in H1299 cells significantly decreased cell viability after paclitaxel or cisplatin treatment in a dose-dependent manner compared with control sgRNA (Figure 3F). Overall, our results suggested that the aberrant upregulation of *PTTG3P* ncRNA in LUAD might perturb the process of mitosis, accelerate cell proliferation, as was predicted, and increase resistance to common LUAD drug treatments in H1299 cells. Our results indicated that the aberrant activation of *PTTG3P* not only contributed to LUAD tumor progression but also decreased drug sensitivity to common LUAD chemotherapeutic drugs, which led to a reduction in therapeutic efficacy.

Knockdown of *PTTG3P* suppressed LUAD tumor growth and prolonged survival time in an orthotopic lung model

To investigate whether *PTTG3P* expression was involved in modulating LUAD tumor growth *in vivo*, we implanted *PTTG3P*-knockdown H1299 cells into the lungs of mice, which served as orthotopic LUAD models, and tracked tumor formation by measuring bioluminescence (BLI) using an IVIS spectrum imaging system. The intensity of luciferase staining as an indicator of tumor growth was detected in the control sgRNA group in 3 weeks after implantation (Supplementary Figure S4A). However, only three mice had traceable BLI signaling in the sg*PTTG3P* knockdown groups, and the majority of the mice in the sg*PTTG3P*-1 and sg*PTTG3P*-2 knockdown groups showed no signs of tumor growth in 6 weeks after implantation (Figure 4A, B; Supplementary Figure S4A). To further interpret the significant tumor suppressive feature, we examine whether *PTTG3P* knockdown could trigger cell death to enhance tumor suppression. Our results indicate that the knockdown of *PTTG3P* enhanced the cleavages of PARP and caspase-3 in H1299 and CL1-5 cells (Supplementary Figure S4B), implying that *PTTG3P* knockdown not only delays cell cycle progression but also induces the cancer cell death led to tumor suppression in an orthotopic lung tumor model.

To validate the association between the upregulation of *PTTG3P* and poor survival in LUAD, we monitored the survival rate of mice for 10 weeks after the implantation of *PTTG3P*-knockdown H1299 cells in orthotopic lung models. Consistent with the minimal tumor growth in the *PTTG3P* sgRNA knockdown groups, the suppression of *PTTG3P* significantly improved the survival of mice after 10 weeks of implantation compared with that of control sgRNA (Figure 4C). We further confirmed *PTTG3P* expression in orthotopic lung tissues using RNA-ISH and hematoxylin and eosin (H&E) staining. Our results showed that higher *PTTG3P* RNA expression was detected in lung tissues of the control sgRNA group compared with expression in the *PTTG3P* knockdown groups (Figure 4D). Overall, our data suggested that *PTTG3P* expression was highly correlated with tumor growth and survival rates in LUAD.

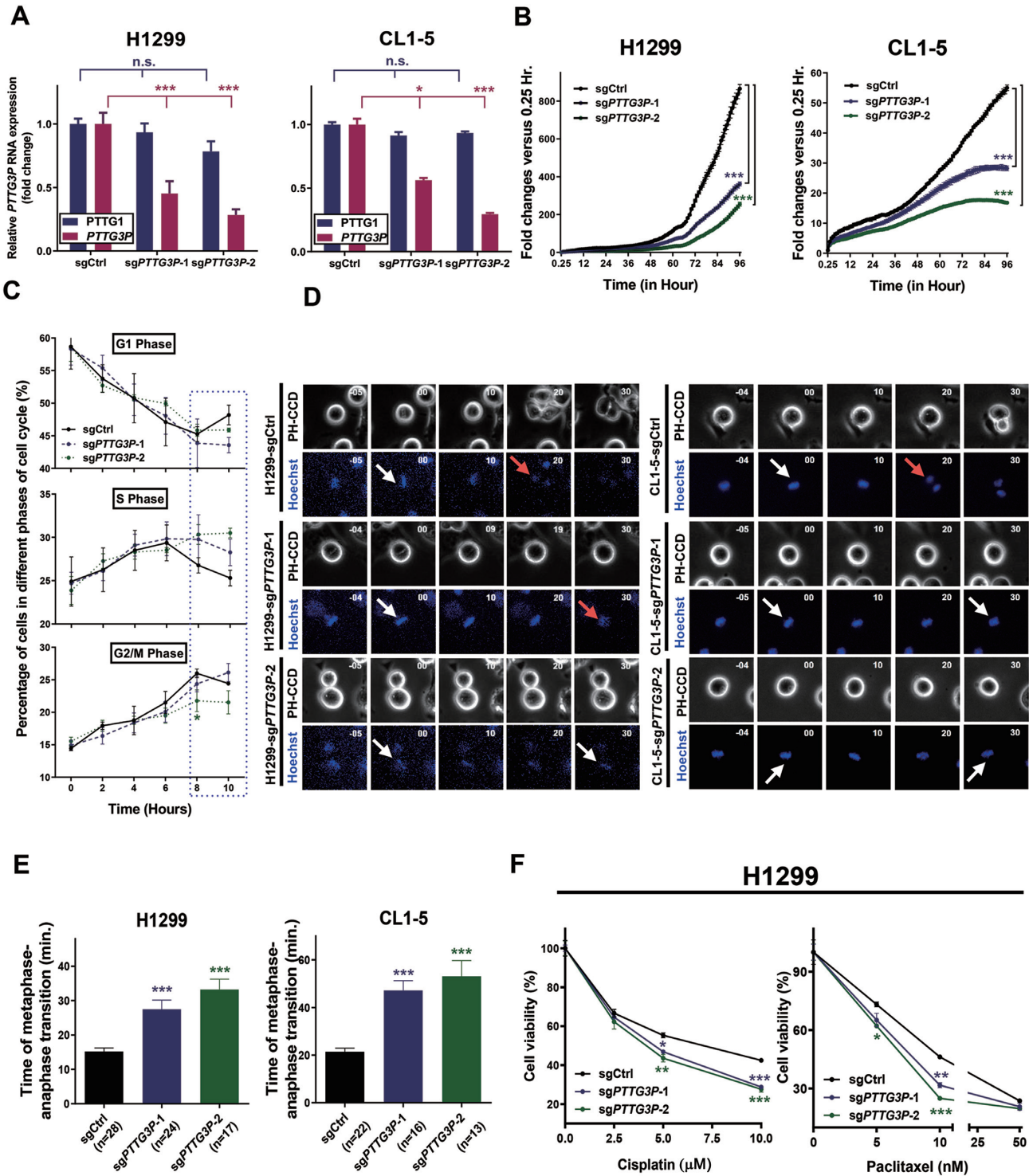


Figure 3. Knockdown of *PTTG3P* with CRISPR/dCas9-KRAB decreased cell proliferation, prolonged metaphase to anaphase transition, and sensitized LUAD cell lines to chemotherapy. (A) Specificity and efficiency of *PTTG3P* RNA knockdown in H1299 and CL1-5 cell lines compared to that of the knockdown of *PTTG1* by with sgRNAs, as determined by a real-time qPCR assay. 18S rRNA served as an internal control; mean±SEM, $n = 3$. (B) *PTTG3P* knockdown in H1299 and CL1-5 cell lines decreased cell proliferation, and the statistics was performed by two-way ANOVA at the final time point. (C) *PTTG3P* knockdown delayed the progression from the G2/M phase to the G1 phase, especially 8 hours after entering the cell cycle, as indicated by the dashed rectangle for the H1299 cell line. (D) *PTTG3P* knockdown prolonged the process of metaphase to anaphase transition when compared to that produced by the control sgRNA, as shown by live-cell time-lapse imaging in H1299 and CL1-5 cell lines. White and red arrows represent metaphase and anaphase, respectively, as defined by Hoechst dye staining. (E) Summary of metaphase to anaphase transition, based on a comparison between the treatment of H1299 and CL1-5 with *PTTG3P* and control sgRNAs, and the statistics was performed by One-way ANOVA with the multiple comparison test. (F) Viability assays of *PTTG3P* and control sgRNA-transfected H1299 cells treated with the LUAD clinical drugs paclitaxel and cisplatin; mean±SEM, $n = 3$; * $P < 0.05$, ** $P < 0.005$, *** $P < 0.0005$.

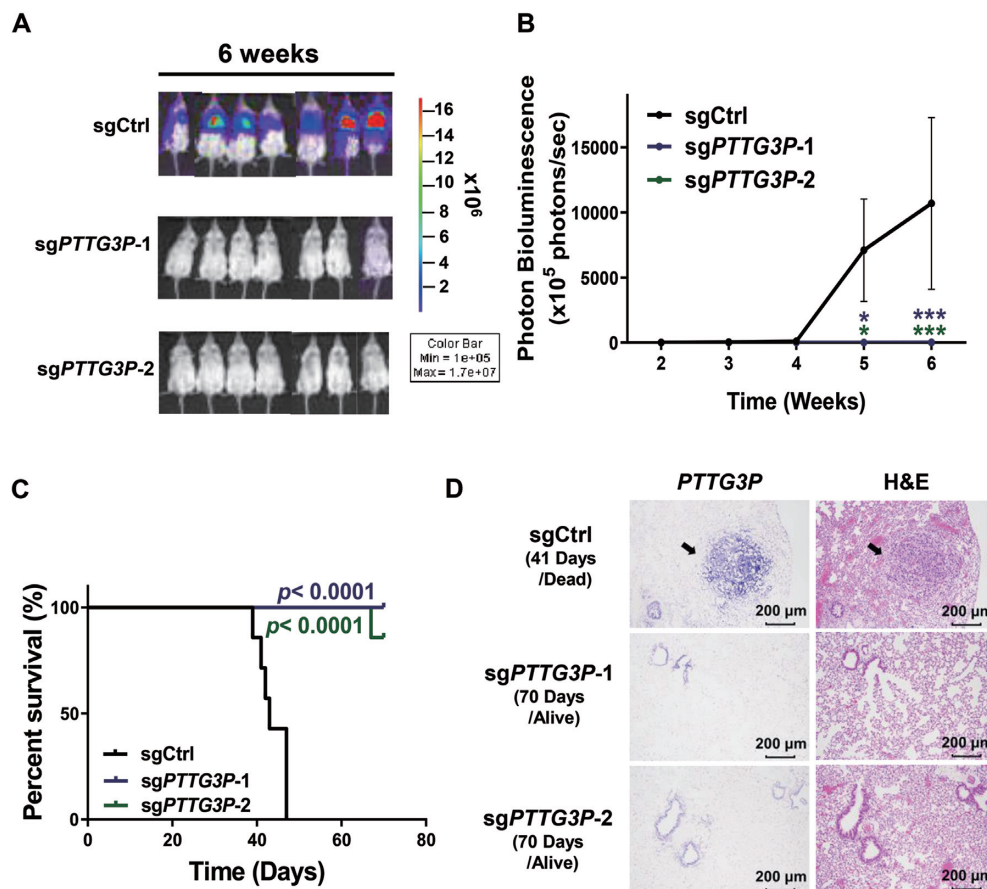


Figure 4. Knockdown of *PTTG3P* inhibited tumor growth and prolonged survival time in an orthotopic model of lung cancer. (A) Representative bioluminescence images of mice bearing 6-week-old tumors inoculated with sgCtrl-Luc⁺ (sgCtrl)- or *PTTG3P*-sgRNA-1/2-Luc⁺ (sgPTTG3P-1/2)-transfected H1299 cells obtained by IVIS; $n = 7$ in each group, mean \pm SEM. (B) Normalized photon bioluminescence images from three groups of mice monitored by IVIS. The error bars show the S.E.M. (C) Kaplan–Meier plot analysis was performed by comparing the survival rates in the three groups of mice. P -values were provided by log-rank tests. (D) Representative images of *PTTG3P* expression during RNA-ISH experiments in tissues from orthotopic lung tumors. *PTTG3P* RNA-ISH and H&E staining were performed on 1-month and 3-month tumor tissues isolated during control and *PTTG3P* knockdown experiments, respectively. * $P < 0.05$, ** $P < 0.005$, *** $P < 0.0005$.

PTTG3P enhances cell proliferation and drug resistance at the RNA level

Because a very low percentage of pseudogenes are transcribed and translated into polypeptides that participate in biological functions (40), we would like to clarify firstly whether the *PTTG3P* pseudogene functions at the RNA and/or protein level as an oncogene to drive LUAD tumorigenesis. Based on polypeptide prediction tools including GenScan (41), CPC (42), iSeeRNA (43) and CPAT (44), *PTTG3P* pseudogene is universally predicted to encode a 22 kDa polypeptide with 202 amino acids in a single exon identical to an unknown protein, Q9NZH4, in the UniProt database (Figure 5A). We established constructs of *PTTG3P* and *PTTG3P*-mu (the ATG codon mutated to the stop codon TAG of the *PTTG3P* predicted translation start site) with HA-tags and conducted assays of *in vitro* cell-free transcription/translation and *in vivo* translation in A549 cells (45). Our results indicated that the *PTTG3P* construct translated a 25 kDa HA-tagged polypeptide but not in the mock and the *PTTG3P*-mu constructs *in vitro* by western blotting analysis (Supplementary Figure S5A). To deter-

mine the essential role of *PTTG3P* polypeptides or RNA in LUAD tumorigenesis, we expressed these constructs in A549 and CL1-0 LUAD cell lines to examine their tumorigenic effects. We found that the RNAs of both *PTTG3P* and *PTTG3P*-mu were overexpressed in comparison with those in mock-transfected cells (Figure 5B) and that *PTTG3P* but not *PTTG3P*-mu expressed a protein detected by the corresponding anti-HA tag antibody that was located at the predicted molecular weight, as shown by western blotting analysis (Figure 5C; Supplementary Figure S5B). The expression of the *PTTG3P* and *PTTG3P*-mu constructs in the A549 and CL1-0 LUAD cell lines induced oncogenic features, including elevated cell proliferation (Figure 5D) and reduced transition time from metaphase to anaphase in mitosis (Figure 5E, F). In contrast to the knockdown of *PTTG3P* in an orthotopic lung tumor model, both overexpressed-*PTTG3P* and *PTTG3P*-mu cells enhanced the tumor growth and reduced the survival rate in comparison to mock in LUAD orthotopic mice models (Figure 5G; Supplementary Figure S5C–E). Furthermore, consistent with the results in knockdown cells treated with cisplatin and paclitaxel, both the overexpressed *PTTG3P* and

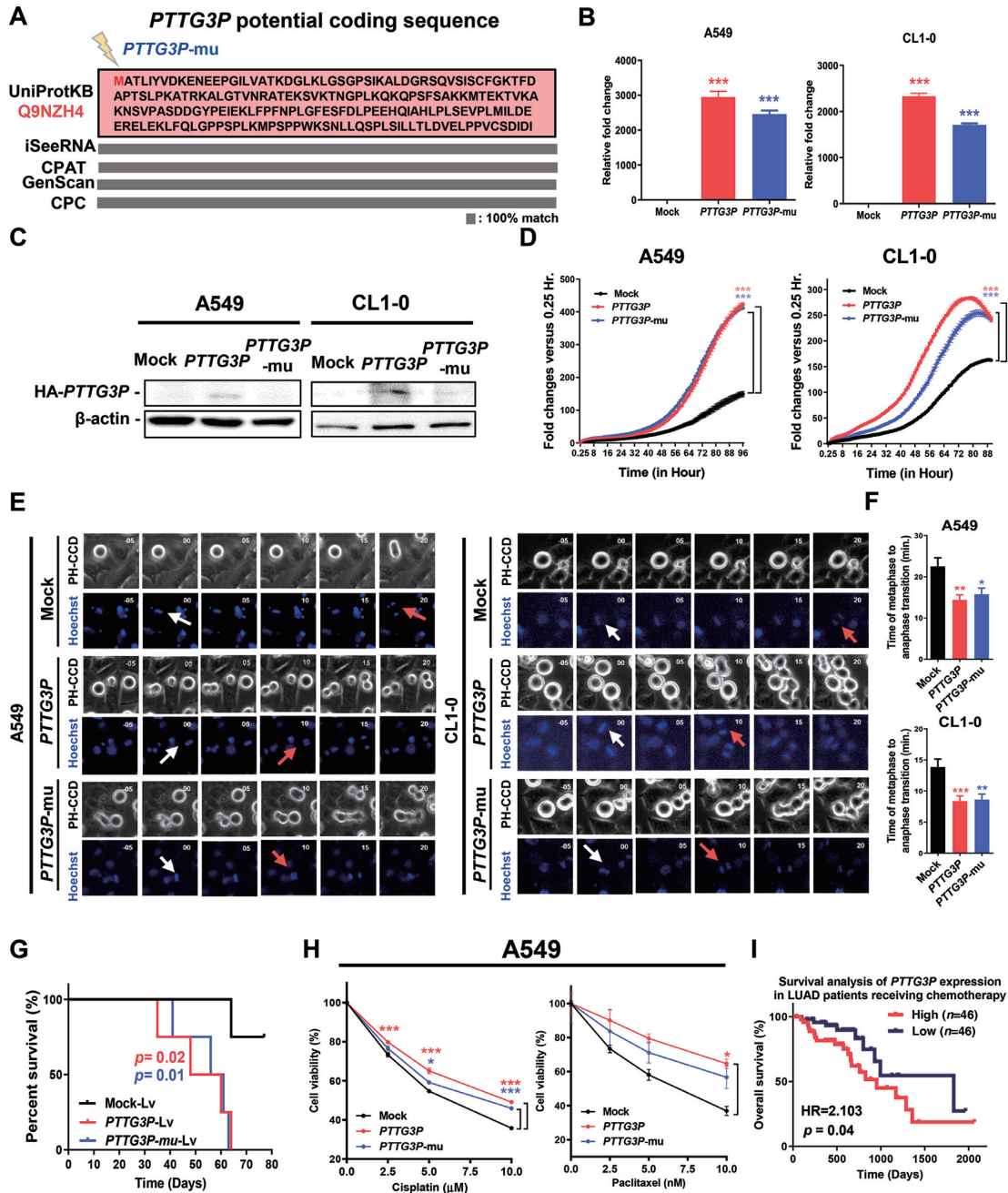


Figure 5. *PTTG3P* ncRNA, but not the corresponding polypeptide, enhanced cell proliferation and drug resistance. (A) The predicted polypeptide sequence of *PTTG3P* produced by multiple prediction tools aligned it with that of the Q9NZH4 protein from the UniProtKB database. The *PTTG3P*-mu construct was generated by mutating the start codon (ATG) into a stop codon (TAG) in the HA-tagged *PTTG3P* construct. (B) RNA expression of *PTTG3P* or *PTTG3P*-mu constructs was detected by real-time qPCR in A549 or CL1-0 cells, and (C) protein expression was detected by western blotting analysis. (D) Ectopic expression of *PTTG3P* increased A549 and CL1-0 cell proliferation, and the statistics was performed by two-way ANOVA at the final time point; mean±SEM, $n = 3$. (E) *PTTG3P* or *PTTG3P*-mu expression shortened the process of metaphase to anaphase in comparison to that observed in the mock-transfected cells, as shown by live-cell time-lapse imaging in A549 and CL1-0 cell lines. White and red arrows represent metaphase and anaphase, respectively, as defined by Hoechst dye staining. (F) Summary of the comparison of the metaphase to anaphase transition in *PTTG3P*-, *PTTG3P*-mu and mock-transfected A549 and CL1-0 cells, and the statistics was performed by One-way ANOVA with the multiple comparison test. (G) Kaplan–Meier plot analysis was performed by comparing the survival rates in the lung orthotopic model bearing the *PTTG3P*-overexpressed A549 cells; Lv: Lentivirus-infected cells; $n = 4$ in each group. P -values were provided by log-rank tests. (H) Expression of *PTTG3P* or *PTTG3P*-mu increased resistance to clinical drugs used in LUAD patients in A549 and CL1-0 cells; mean±SEM, $n = 3$. (I) Kaplan–Meier plot of data from LUAD patients treated with chemotherapy in the TCGA cohort. The survival analysis was examined with the Gehan-Breslow–Wilcoxon test with tertile criteria. * $P < 0.05$, ** $P < 0.005$, *** $P < 0.0005$.

PTTG3P-mu cells showed increased cell viability during treatment with cisplatin and paclitaxel in a dose-dependent manner (Figure 5H). These results suggested that *PTTG3P* ncRNA *per se*, regardless of its translation into a polypeptide, is oncogenic in LUAD tumorigenesis. Moreover, we found that the high expression of *PTTG3P* in LUAD patients in the TCGA cohort resulted in a poor survival rate during treatment with chemotherapy, in contrast to the improved survival rate in LUAD patients with low *PTTG3P* expression (Figure 5I). Collectively, our results implied that the upregulation of *PTTG3P* ncRNA but not the predicted polypeptide in LUAD cells could induce oncogenesis and resistance to common LUAD drug treatments that might lead to an association with poor patient prognosis.

***PTTG3P*, as a chromatin-associated ncRNA, increased cell proliferation and drug resistance in LUAD cells by enhancing FOXM1-BUB1B oncogenic signaling**

To uncover the oncogenic mechanisms of *PTTG3P* ncRNA in LUAD, we performed RNA sequencing (RNA-seq) in *PTTG3P*-overexpressing A549 and *PTTG3P*-knockdown H1299 cells and their corresponding control cells, followed by gene enrichment analyses (GSE114826). Because of the high consistency of the two transcriptomic datasets, we uncovered 1543 differentially expressed genes (DEGs) and conducted hallmark pathway analysis with GSEA and drug association analysis with WebGestalt (Figure 6A, upper panel; Supplementary Table S7). Consistent with the prediction shown in Figure 1, not only were 1543 DEGs significantly enriched in the cell cycle-associated gene lists, especially in hallmark pathways involved in the mitotic-spindle and G2/M checkpoints, but these DEGs also showed a strong association with the chemotherapeutic drugs cisplatin (Figure 6A, lower panel) and paclitaxel (P -value = 0.0145).

To identify the *PTTG3P*-mediated downstream mitosis-related genes, we examined the association of the 1543 DEGs with the genes in the aforementioned *PTTG3P*-associated networks in each LUAD transcriptome dataset. We found that BUB1B showed the most consistent association in the *PTTG3P*-associated networks and in the LUAD transcriptomes which was upregulated in *PTTG3P*-overexpressing cells and downregulated in *PTTG3P* knockdown cells (Figure 6B). Interestingly, BUB1B is a mitotic checkpoint protein that plays a critical role in regulating the spindle-assembly checkpoint and that shows aberrant expression in several cancers (46). Therefore, we selected BUB1B as a possible *PTTG3P* downstream target that may contribute to aberrant mitosis and poor prognosis.

We found that the expression of *PTTG3P* in A549 cells upregulated BUB1B expression at both the RNA and protein levels (Figure 6C, D). In contrast, the knockdown of *PTTG3P* expression in H1299 cells reduced BUB1B expression in terms of its RNA and protein levels. To confirm that BUB1B is the downstream target of *PTTG3P*, we compared the impacts of cell viability by conducting the *PTTG3P* knockdown (*sgPTTG3P*) with BUB1B overexpression in H1299 cells and *PTTG3P* overexpression with BUB1B knockdown (shBUB1B) in A549 cells (Supplementary Figure S6A). Interestingly, cell proliferation was

slightly rescued by BUB1B overexpression upon *PTTG3P* depletion, conversely, the overexpression of *PTTG3P* failed to rescue the cessation of cell proliferation under BUB1B depletion (Figure 6E). Moreover, the cell resistance to cisplatin and paclitaxel showed a similar trend with the effect of cell proliferation (Figure 6F). Together, we concluded that BUB1B is the downstream of *PTTG3P*-driven LUAD tumorigenesis.

To clarify how *PTTG3P* regulated BUB1B, we examined the subcellular localization of *PTTG3P* since the localization of the ncRNA might affect RNA biological functions (47,48). Our results showed that *PTTG3P* is more highly expressed in chromatin than in the cytoplasm and nucleoplasm fractions, as detected by RNA fluorescence *in situ* hybridization (RNA-FISH) and RNA subcellular localization (Figure 6G). Because of the likely role played by nuclear chromatin ncRNAs in the modulation of gene transcription, we speculated that the upregulation of *PTTG3P* ncRNA might facilitate mitotic catastrophe via regulating BUB1B transcriptional activation by modulating its promoter region (49). Indeed, the expression of *PTTG3P* directly modulated the transcription of BUB1B by targeting the (−350/−216) promoter region of BUB1B, which was not observed in the mock control, as shown by luciferase reporter assays (Figure 6H).

Next, we speculated that FOXM1 is one of the transcription factors that binds to the (−350/−216) promoter region of BUB1B along with *PTTG3P* ncRNA because the FOXM1/BUB1B signaling pathway is essential for the tumorigenicity of glioblastoma and rhabdomyosarcoma, especially in terms of mitotic progression (50,51). First, we examined BUB1B expression during FOXM1 knockdown in *PTTG3P*-overexpressing cells (Supplementary Figure S6B) and found that BUB1B expression was significantly decreased, suggesting that *PTTG3P* might modulate BUB1B expression via FOXM1. Second, we examined whether upregulated *PTTG3P* ncRNA might interact with FOXM1 to target the BUB1B promoter to affect transcriptional activation during aberrant mitosis in tumorigenesis. Indeed, the expression of *PTTG3P* potentiated FOXM1 binding to the BUB1B promoter region in A549 cells as detected by chromatin immunoprecipitation (ChIP) assays (Figure 6I).

Third, we revealed that sense- but not antisense-strand *PTTG3P* interacted with FOXM1 with an RNA pull-down assay (Figure 6J). In addition, FOXM1 was shown to interact with *PTTG3P* and was also detected by an RNA immunoprecipitation (RIP) assay (Figure 6K), strongly suggesting that the upregulated *PTTG3P* ncRNA interacted with the FOXM1 transcription factor to target the BUB1B promoter, which resulted in aberrant transcriptional activation and facilitated mitosis. Finally, we performed chromatin isolation by using RNA purification (ChIRP) to confirm that *PTTG3P* ncRNA and FOXM1 form a complex on the BUB1B promoter with 3-fold stronger intensity than the LacZ control (Figure 6L; Supplementary Figure S6C, D). Collectively, we established a systematic approach to reveal prognostic ncRNAs networks that could predict potential biological functions and pathological mechanisms involved in LUAD. Moreover, we identified *PTTG3P* as a functional ncRNA that guides and complexes with FOXM1 on the BUB1B promoter to augment BUB1B transcriptional activ-

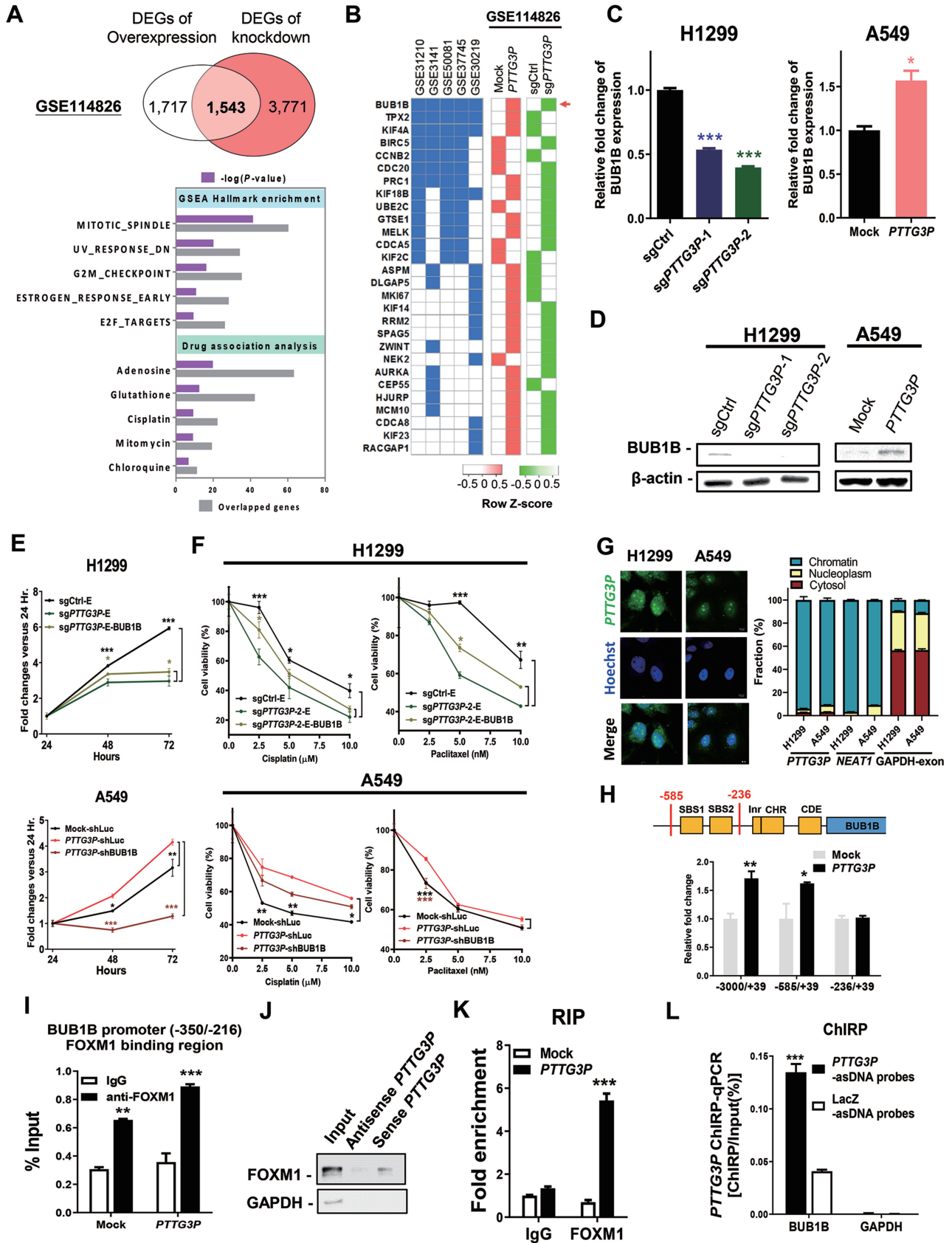


Figure 6. Interaction of *PTTG3P* ncRNA with FOXM1 augments BUB1B transcriptional activity to enhance cell proliferation and drug resistance. (A) 1543 DEGs were shared between the upregulated genes in *PTTG3P*-overexpressing A549 cells and the downregulated genes in *PTTG3P*-knockdown H1299

ities to facilitate mitosis, tumor growth and drug resistance, resulting in poor prognosis in LUAD patients.

DISCUSSION

To systematically determine the biological functions, clinical relevance and therapeutic targets of ncRNAs, we established a pipeline for the identification of prognostic ncRNAs and their associated networks. The prognostic ncRNA-containing co-expressed gene module allowed us to produce precise predictions of the functions and drug responses of ncRNAs with gene enrichment analysis. By performing validation in multiple LUAD cohorts, we identified the upregulation of *PTTG3P* ncRNA in LUAD and revealed its involvement in functional mechanisms associated with drug sensitivity and prognosis. We demonstrated that *PTTG3P* ncRNA interacts with FOXM1 to target the downstream BUB1B promoter, leading to aberrant transcriptional activation that facilitates the mitotic metaphase to anaphase transition and tumor growth and results in poor prognosis in LUAD animal models and human patients. Moreover, LUAD cells with *PTTG3P* upregulation acquired resistance to treatment with paclitaxel and cisplatin in cell models and in LUAD patients from TCGA cohorts, which further suggested that the detection of up-regulated *PTTG3P* ncRNA might be crucial for therapeutic intervention in LUAD patients.

Several strategies were applied for the functional prediction of ncRNAs, including the prediction of ncRNA-protein interactions, the determination of subcellular localization, the analysis of disease gene clustering and networks, and the analysis of the enrichment of transcriptional factor targets to reveal the downstream signaling pathways of ncRNAs (52–54). In lung cancers, several prognostic coding and non-coding RNAs were identified based on different bioinformatic approaches, especially focusing on prognostic signature gene sets for predicting patient's therapeutic responses and survival at the early stage (9,30,55–57). With increasing evidence of ncRNAs participating in tumor progression, more efforts conducted for exploring their prognostic value in LUAD. For instance, the lncRNA-mRNA-miRNA competing endogenous RNA networks from LUAD differential expressed RNAs were established to reveal prognostic RNAs in association with LUAD patient survival (58,59). Two lncRNAs (*DKFZP434 L187* and

LOC285548) were discovered for their prognostic properties in LUAD based on their differential expression (10), and another two lncRNA (*C1orf132* and *TMPO-AS1*) were identified to predict survival of stage I-II LUAD patients without adjuvant therapies (60). We noticed that these predicted prognostic LUAD coding genes and ncRNAs based on transcriptome analysis are not identical owing to heterogeneity of LUAD samples and analytic algorithms. Additional endeavors on functional validations and mechanistic actions of identified ncRNAs in LUAD should be critical for revealing useful prognostic ncRNAs to improve LUAD therapeutic interventions.

In comparison to other similar approaches (55–57,61), several features of our study are important for identifying robust prognostic ncRNAs in heterogeneous LUAD transcriptomes that are suitable for functional prediction in other cancer types. First, we collected more than 604 LUAD transcriptomic datasets along with 10-year survival information from different ethnic groups to reveal the common etiology in LUAD (Supplementary Table S1). Second, we applied highly stringent criteria to the already restricted prognostic noncoding gene networks, including requirements for highly correlated genes with cutoff heights of 0.95 and minimal module sizes of 5 genes in multiple datasets, to obtain promising prognostic ncRNAs. Finally, we conducted comprehensive *in silico*, pathological and experimental validation with mechanistic understanding in mouse cancer models and human patients to obtain ncRNAs that could be used as innovative LUAD therapeutic markers.

In addition to *PTTG3P* ncRNA in LUAD, the remaining five prognostic ncRNAs including *MIR497HG*, *HSP078*, *TBX5-AS1*, *LOC100506990*, *C14orf64* and their WGCNA networks were predicted to participate in biological functions and pathways involved in developmental biology, axon guidance, the nicotine degradation II pathway, Slit/Robo signaling and Sema4D- and integrin-related cell migration during progression in LUAD patients. Except for *MIR497HG*, the remaining 4 ncRNAs *HSP078* (*LOC101928517*), *TBX5-AS1*, *LOC100506990* and *C14orf64* (*LINC01550*) have been rarely studied in cancers. One of the mature forms of the *MIR497HG* cluster host gene, *hsa-miR-195*, was shown previously to function as a negative regulator of lung cancer growth and metas-

cells (GSE114826), which were further analyzed with GSEA and WebGestalt. (B) Genes in the *PTTG3P*-associated networks in five LUAD datasets that were ranked according to their *k*Within value from the WGCNA were aligned with the common DEGs in the deposited transcriptomes in GSE114826 dataset. The GSE19188 dataset was excluded because no *PTTG3P*-associated module was constructed. Heat maps were individually generated on the basis of the Z-score by Heatmapper and show the upregulated (red) and downregulated (green) genes. A red arrow indicates the highest concordant gene in each dataset. (C) BUB1B expression was examined at the RNA level by real-time qPCR, and (D) its protein level was determined by western blotting analysis in H1299 and A549 cells. (E) Proliferation assays and (F) cell viability assays were performed after treatment with clinical drugs in *PTTG3P* knockdown H1299 cells overexpressing BUB1B with an EGFP-tag (*sgPTTG3P-E-BUB1B*) or in *PTTG3P*-overexpressing A549 cells with shBUB1B knockdown (*PTTG3P-shBUB1B*). (G) The cellular distribution of *PTTG3P* ncRNA detected by fluorescence RNA-ISH and the RNA expression of *PTTG3P* in the cytoplasmic, nucleoplasmic, and chromatin fractions were examined by real-time qPCR in both of H1299 and A549 cell lines. GAPDH-exon and *NEAT1* RT-qPCR amplicons served as RNA markers for the cellular and nuclear fractions, respectively. (H) The transcriptional activities of different promoter regions of BUB1B were detected in A549 transfectants with luciferase reporter assays. (I) The binding region of *PTTG3P* ncRNA/FOXM1 versus FOXM1 alone on the BUB1B promoter was determined by a chromatin immunoprecipitation assay with an anti-FOXM1 antibody and real-time qPCR. (J) Western blotting analysis of FOXM1 obtained from the RNA pull-down assay was performed in nocodazole-synchronized H1299 cells. (K) *PTTG3P* ncRNA was immunoprecipitated with an anti-FOXM1 antibody and examined by real-time qPCR for an RNA-IP assay. (L) Chromatin isolation by RNA purification (ChIRP) in the presence of FOXM1 was conducted with an antisense *PTTG3P* DNA probe (*PTTG3P-asDNA* probe) and a LacZ antisense probe (LacZ-asDNA probe) for a comparison of their binding to the promoters of BUB1B and GAPDH with a ChIRP-qPCR assay. GAPDH and a LacZ-asDNA probe served as negative controls; mean \pm SEM, $n = 3$. * $P < 0.05$, ** $P < 0.005$, *** $P < 0.0005$.

tasis by targeting the oncogenic *CHEK1*, *HDGF*, *IGF1R*, *MYB* genes and other genes, which gave it prognostic value (62–66). Interestingly, *hsa-miR-195* was found to be one of the core schizophrenia genes participating in axonal guidance signaling that is consistent with our WGCNA and functional prediction (67). Notably, several predicted biological pathways involved in axon guidance, the Slit-Robo pathway, sema4D-related and integrin-related cell migration, and developmental biology that were identified with these prognostic ncRNA networks were previously shown to contribute to lung cancer tumor progression and to be related to guidance molecules in the nervous system that control the direction of axon outgrowth (33). Owing to the relatively small number of overlapping probes, the level of statistical significance was relatively low, therefore further exploration of the pathological mechanisms of guidance molecules in LUAD progression is needed to warrant the development of novel therapeutic markers.

Our results also clarify that *PTTG3P* functions as a FOXM1-interacting ncRNA to target and activate the mitotic checkpoint kinase BUB1B without the need to translate the predicted polypeptide, leading to *PTTG3P*-mediated LUAD tumorigenesis and drug resistance. Notably, *PTTG3P* is labeled an ‘uncertain protein’ in the UniProt database, and no predicted peptide fragments of *PTTG3P* were detected in the database of the human proteome map (www.humanproteomemap.org). These indications further support our results that *PTTG3P* functions as a ncRNA. Moreover, we also found that upregulated *PTTG3P* is associated with patients with LUAD that is classified as the AD.1 subtype, which is a genomic and molecular subtyping group of LUAD based on the clustering of multiomics data in TCGA (68) (Supplementary Figure S7). Consistently, the AD.1 subtype in LUAD patients conferred poor differentiation and malignant tumor features based on the enrichment of cell-cycle associated genes, leading to a poor prognosis in LUAD. Furthermore, our analysis of aberrant *PTTG3P* expression in 36 cancer types in TCGA datasets indicated that *PTTG3P* is upregulated in more than half of the cancer types examined, and 10 out of 36 cancer types, including LUAD, exhibited a high and statistically significant hazard ratio for patient survival (Supplementary Figure S8). Overall, all of these results provide lines of evidence that show that the upregulation of *PTTG3P* ncRNA at minimum activates FOXM1/BUB1B oncogenic signaling to facilitate cancer cell proliferation and tumor growth along with malignant progression, leading to poor prognosis in cancer patients. In conclusion, our study provides a highly stringent pipeline that can be used to reveal prognostic ncRNAs with weighted co-expression networks for accurate mechanical prediction and the determination of their association with drug responses. Moreover, we demonstrated that the biological function and pathological mechanisms involved in the upregulation of the *PTTG3P* ncRNA/ FOXM1/BUB1B axis resulted in poor prognosis in LUAD patients. Notably, the activation of *PTTG3P* ncRNA during cancer progression increased the resistance of LUAD cells to paclitaxel and cisplatin treatment, and validation in TCGA patients who received chemotherapy might reveal an important biomarker that could be used for dosage determination when evaluating the optimal thera-

peutic intervention for LUAD. Importantly, our approach would be suitable for use in other cancer types to develop new prognostic ncRNAs as novel therapeutic targets to benefit cancer patients.

DATA AVAILABILITY

To review GEO accession GSE114826: Go to <https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE114826>.

SUPPLEMENTARY DATA

Supplementary Data are available at NAR Online.

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