Original Research

ARNTL2 is an indicator of poor prognosis, promotes epithelial-to-mesenchymal transition and inhibits ferroptosis in lung adenocarcinoma

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A B S T R A C T

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Objectives: ARNTL2, as a circadian transcription factor, has been recently proposed to play an important role in a variety of tumors. However, the role of ARNTL2 in lung carcinogenesis and progression remains unclear. The purpose of this study was to investigate the effect of ARNTL2 on the clinical characteristics and prognosis of lung adenocarcinoma and to explore the relationship between ARNTL2 and EMT, ferroptosis in lung adenocarcinoma.

Methods: The Cancer Genome Atlas (TCGA) database’s multi-omics data were downloaded using the Xena browser. Based on the expression levels of ARNTL2, patients with lung adenocarcinoma from TCGA were divided into two groups: those with high ARNTL2 expression and those with low ARNTL2 expression. ARNTL2 was studied for its effects on lung adenocarcinoma’s clinicopathological, genomic, and immunological characteristics. Furthermore, in vivo and in vitro assays were used to confirm the impact of ARNTL2 knockdown on lung adenocarcinoma cell function.

Results: We found ARNTL2 is highly expressed in lung adenocarcinoma and was an independent predictor of a poor prognosis in patients with lung adenocarcinoma. In addition, we demonstrated that knockdown of ARNTL2 promoted ferroptosis, inhibited EMT, cell proliferation, migration and invasion in lung adenocarcinoma. In contrast, overexpressing ARNTL2 yielded the opposite results.

Conclusions: ARNTL2 is an independent unfavorable prognostic factor for lung adenocarcinoma. It plays a facilitating role in the development of lung adenocarcinoma, especially in promoting EMT and inhibiting ferroptosis, revealing that ARNTL2 may be a potential biomarker for lung adenocarcinoma.

Introduction

In 2020, lung cancer has been the leading cause of cancer death worldwide [1]. The most common subtype is lung adenocarcinoma (LUAD) [2]. Despite efforts and advancements in lung adenocarcinoma therapy, the prognosis of LUAD patients remains bleak; the 5-year survival rate of lung cancer patients was less than 20% [3]. As a result, it is critical to investigate the etiology and mechanisms of LUAD malignant progression to develop more effective treatment strategies.

Circadian rhythms are oscillations that occur every 24 h and can...
initiate and monitor various biological processes [4]. Circadian rhythms in mammals are controlled by a central clock in the suprachiasmatic nucleus of the anterior hypothalamus and peripheral oscillators found in many peripheral tissues [5]. Many studies have demonstrated that environmental disruptions to the circadian rhythm may increase the risk of certain cancers [6]. A critical circadian transcription factor Aryl Hydrocarbon Receptor Nuclear Translocator Like 2 (ARNTL2), has been shown to participate in the organization of feedback loops to provide near 24 h rhythmicity [7]. It has also been demonstrated to play an essential role in various tumors. ARNTL2 is associated with poor survival and immune infiltration in clear cell renal cell carcinoma [8] and triple-Negative Breast Cancer [9]. Besides, ARNTL2 has also been found to promote metastasis of colon carcinoma [10] and estrogen Receptor-Negative Breast Cancer [11]. However, research on the role of ARNTL2 in lung adenocarcinoma is scarce and insufficiently thorough. In our study, we first found that ARNTL2 promoted EMT and inhibited ferroptosis in LUAD cells.

The current study analyzed 513 lung adenocarcinoma samples from the Cancer Genome Atlas (TCGA database); we discovered that ARNTL2 is highly expressed in lung adenocarcinoma and has a negative impact on LUAD patient’s prognosis. Furthermore, ARNTL2 expression was found to significantly impact the progression and immune microenvironment of lung adenocarcinoma. Through in vivo and in vitro experiments, we also confirmed that ARNTL2 promotes the proliferation, migration, and invasion of lung adenocarcinoma cells. The findings suggest that ARNTL2 plays a vital role in developing, progressing, and treating lung cancer, laying the groundwork for future mechanistic research and treatment.

Methods

Data processing

Gene expression data of LUAD patients (FPKM normalized) and corresponding clinical and survival information of The Cancer Genome Atlas (TCGA) were downloaded from the UCSC Xena browser (GDC hub: https://xenahubs.net) [12]. The data with missing prognosis information, including outcome status and survival time, was removed. The edition of lung cancer staging was classified according to the American Joint Committee on Cancer (AJCC) TNM Classification for Lung and Pleural Tumors (eighth edition).

TIMER analysis

The “DiffExp” module in TIMER [1] web server [13] was applied in examining the expression level and matched non-carcinoma tissues in LUAD tumors.

Genome statistical analysis

Based on the median values of ARNTL2 expression level, TCGA cases were divided into high and low groups. The following genome statistical analyses were carried out in R version 4.0.3:

(a) Differentially expressed genes (DEGs): DEGs and miRNAs were identified using the limma package. To calculate DEGs and miRNA expression changes, the moderated t-test was used. The Benjamini and Hochberg method was used to adjust the P-value as FDR. The adjusted P-value (0.05, and the log fold change) > 0.5 [14].

(b) Copy number variations (CNV) and microRNAs (miRNAs): the maftools package was used to compare the distribution of somatic mutations and the types of CNV [15]. The significance of the mutational frequency was determined using an adjusted P-value < 0.01. The Kruskal-Wallis test was used to compare the somatic mutations and types of copy number variations between the high and low-ARNTL2 groups, and an adjusted P-value < 0.05 was considered statistically significant. The oncoplot function was used to display the results.

(c) GO and KEGG: The clusterProfiler R package was used to perform GO, and KEGG pathway enrichment analysis on the DEGs’ involved pathways and biological functions [14]. The cut-off values were determined to be adjusted P < 0.05 and false discovery rate (FDR) < 0.05.

(d) Ferroptosis analysis: Ferroptosis-related genes are derived from Ze-Xian Liu et al.’s systematic analysis of Ferroptosis’s aberrances and functional implications in Cancer [16].

(e) miRNA analysis: To calculate mRNAs, we use the OCLR algorithm developed by Malia et al. [17]. The gene expression profile contains 11,774 genes based on mRNA expression characteristics. We used the same Spearman correlation (RNA expression data), subtracted the minimum value, and divided it by the maximum value’s linear transformation, which maps the dryness index to the range [0, 1].

(f) Drug sensitivity analysis: Using the largest publicly available pharmacogenomics database [the Genomics of Drug Sensitivity in Cancer (GDSC), https://www.cancerrxgene.org/], we predicted the chemotherapeutic response for each sample. The R package “p8Prophetic” was used to implement the prediction process, in which the samples’ half-maximal inhibitory concentration (IC50) was estimated using ridge regression, and the prediction accuracy was calculated. The default values were used to set all parameters after removing the batch effect of combat and tissue type from allSoldTumours, and duplicate gene expression was summarised as a mean value [18].

(g) Immune infiltration and immune checkpoints: To estimate immune infiltration, we used the QUANTISEQ method from the Immunodeconv R package [19]. SIGLEC15, TIGIT, CD274, HAVCR2, PDCD1, CTLA4, LAG3, and PDCD1LG2 were chosen as immune-checkpoint-relevant transcripts and their expression levels were determined. Based on the RNA-seq expression profiles, the QUANTISEQ method (https://quantuseq.stanford.edu) [20] was chosen to analyze the relative levels of the ten tumor-infiltrating immune cell phenotypes. The distribution of ten tumor-infiltrating immune cell types between poor and good prognostic signatures was studied.

(h) Correlation analysis of ARNTL2 expression and TMB [21,22]:

We used Spearman’s correlation analysis to describe the correlation between quantitative variables that did not have a normal distribution. The Pearson correlation coefficient has a value range of [-1,1]. A higher absolute value indicates a stronger association, and the sign indicates whether the two variables are associated positively or negatively. The density curve on the right represents the TMB score, while the upper-density curve represents the gene’s distribution trend.

The R foundation implemented all aforementioned analysis methods and the R package for statistical computing (2020) version 4.0.3.

Single-cell analysis

Single-cell analysis was performed through the CANCERSEA website. CancerSEA is the first dedicated database that comprehensively decodes distinct functional states of cancer cells at single-cell resolution [23].

Prognosis analysis

Prognostic information was analyzed in the PrognoScan and Kaplan-Meier plotter database [24].
Lung adenocarcinoma cell lines (A549 and H1299) were purchased from the Chinese Academy of Sciences Cell Bank. Cells were cultured in DMEM (Hyclone, Logan, UT, USA) supplemented with 10% fetal bovine serum (Hyclone), 100 U/mL penicillin, and 100 U/mL streptomycin in a humidified 5% CO2 atmosphere at 37 °C.

Two different short hairpin RNAs (shRNAs) encoding shARNTL2 and scramble shRNA control (catalog) was designed by Shanghai Genechem Co., Ltd, and cloned into a lentivirus vector with puromycin resistance, which was then transfected into cells and screened with puromycin to ensure transfection efficacy. Viral transduction was performed as the manufacturer’s protocol. After three days of virus transfection, knock-down was verified by western blot analysis using ARNTL2-specific antibody (ab221557, Abcam). Sequences of the shRNAs and the control are provided in Supplementary Table 1.

Compounds

The following compounds were obtained from Topscience (USA): RSL3 (T3646), cisplatin (T1564). RSL3 was dissolved in PBS (Beyotime, Shanghai, China) and cisplatin in DMSO (Beyotime), according to their solubility, and then stored at −20 °C.

Western blot

RIPA buffer (Beyotime, Shanghai, China) containing protease and phosphatase inhibitors cocktail (Beyotime) were used to extract proteins, enhanced BCA Protein Assay Kit (Beyotime) was used for protein quantification. As previously reported [25], Protein samples were stained with 4% formaldehyde/0.005% gentian violet solution and incubated for 12 h at 4 °C. Then, tris-buffered saline Tween-20 (TBST) was used to wash the membranes three times. After washing, the membranes were incubated with secondary antibodies at room temperature for one hour. Finally, the protein bands were visualized and analyzed using a Moon Chemiluminescence Reagent kit (Beyotime). Three biological replicates were analyzed and western blots repeated three times. The following antibodies were used in this study: Anti-ARNTL2 antibody (1:1000, Abcam, ab221557), Anti-E-Cadherin antibody (1:1000, Cell Signaling Technology, #9782), Anti-N-Cadherin antibody (1:1000, Cell Signaling Technology, #9782), Anti-β-catenin antibody (1:1000, Cell Signaling Technology, #9782), Anti-Vimentin antibody (1:1000, Cell Signaling Technology, #9782), Anti-Snail antibody (1:1000, Cell Signaling Technology, #9782), Anti-Fibronectin (1:1000, Abcam, ab2413), Anti-NF-E2L2 antibody (1:1000, Cell Signaling Technology, #12,721), Anti-GS2 antibody (1:1000, abcam, ab13509), Anti-SLC7A11 antibody (1:1000, Cell Signaling Technology, #98,051), Anti-CISD1 antibody (1:1000, Cell Signaling Technology, #83,775), and Anti-β-Actin antibody (1:1000, Beyotime, AF0003).

Cell viability assays

For cell proliferation assays, cells were seeded in quintuplicate into black 96-well plate at a density of 2000 cells per well at logarithmic growth phase. After incubation at 37 °C for 0, 24, 48, 72, 96, and 120 h. For cytotoxicity assays, 5000 cells per well were seeded in quintuplicate in 96-well plates and incubated for 24 h. Cells were treated with different doses of chemotherapy drugs for specific times as needed. The cell proliferation was measured by Enhanced Cell Counting Kit-8 Viability Assay Kit (Beyotime) according to the manufacturer’s protocol.

Wound healing assay

Control shRNA and shARNTL2 cells were inoculated on the 6-well plate on average. Cells were grown into monolayer and manual scratching with a 200 µl pipette tip. Cells were rinsed with PBS and incubated at 37 °C in serum-free media. Photographs of the wounded areas were taken every 24 h by phase-contrast microscopy.

Transwell migration and invasion assay

Cell migration and invasion abilities were measured by Transwell assays using the 24-well transwell chambers with 8 µm polycarbonate membranes (Corning, NY). The uncoated were used to determine migration, and pre-coated with Matrigel Basement Membrane Matrix were used to determine invasion (BD Biosciences). The chambers were rehydrated in a serum-free medium for 2 h as described by the manufacturer. Then the upper chambers were added with serum-free medium, while the lower chambers were added with serum medium. Cells were seeded onto the upper chambers with a density of 5 × 104 cells per well and incubated for 24 h at 37 °C, 5% CO2. Cells migrated toward the lower chambers were fixed with methanol and stained with 0.5% crystal violet. Each assay was photographed under the inverted microscope (Olympus), and the number of cells within each chamber was counted by ImageJ software.

Clone formation assay

Control or ARNTL2 shRNA-transduced A549 and H1299 cells (3 × 103 cells/well) were cultured in 6-well plates were cultured at 37 °C in 5% CO2 environment. ARNTL2 shRNA-transduced or control cells were seeded in 6-well plates at a density of 5 × 106 cells per well and cultured at 37 °C in 5% CO2 environment. After nine days, cells were stained with 4% formaldehyde/0.005% gentian violet solution and captured under the inverted microscope.

Immunohistochemistry and immunofluorescence

The tissue specimens were collected from both tumor and adjacent normal tissues of 100 patients with LUAD who received surgery from September to November 2020 in the Zhongshan Hospital. As previously reported [26], the tissues were stained by a GTVision + Detection System/MoRb Immunohistochemistry kit (GK500710, GeneTech, Shanghai, China) following the manufacturer’s protocol. Specifically, the 5-µm paraffin-embedded tissues were dewaxed, rehydrated, and incubated with antibodies against ARNTL2 (1:500, Abcam, Cambridge, UK) at 4 °C overnight, and then were incubated with biotinylated secondary antibodies. For immunofluorescence, sections were incubated with primary antibodies against ARNTL2 (rabbit polyclonal, 1:500), followed by incubation with the respective secondary antibodies (Cy3-labeled goat anti-rabbit IgG). DAPI nuclear counterstaining was then performed. Finally, a fluorescence microscope was used to take micrographs.

Subcutaneous tumor formation

Female BALB/c nude mice (4 weeks old) were obtained from Gem-Pharmatech Co., Ltd, (Jiangsu, China) and housed and maintained in laminar airflow cabinets under specific pathogen-free conditions. For the subcutaneous A549 tumor model, 1 × 106 A549 cells (Group 1, sh-control; Group 2, sh1-ARNTL2) were injected subcutaneously into the right flank of each mouse (5/group). In life, mice were monitored regularly, and tumors were measured with vernier calipers every 5 days (tumor volume was calculated as π/6 × length × width × width). The mice were euthanized after 6 weeks, and tumors were isolated and weighed for further analyses.
Statistical analysis

All experiments were performed in at least triplicates. Statistical analyses in the current study were completed by R version 3.6.1 (R Foundation for Statistical Computing, Vienna, Austria). The R package included survival, rms, ggplot2, limma, maftools, clusterProfiler, and ImmunoCon. Statistical significance was set at a two-sided \( P \)-value < 0.05. Categorical variables were compared using Fisher’s exact test. Pearson’s \( r^2 \) test and continuous variables were compared using Student’s \( t \)-test and the Wilcoxon test. Multivariable Cox regression analyses were used to test independent prognostic value using the \( R \) package survival and the coxph function. Image analysis was performed using ImageJ software.

Result

Gene expression and clinical features

An earlier study discovered that ARNTL2 was frequently overexpressed in various cancers [27]. Fig. 1A depicts the differential expression of ARNTL2 between tumor and adjacent normal tissues in TCGA. To further validate whether ARNTL2 was differentially expressed in lung adenocarcinoma tissues, 513 cases of LUAD patients’ data from TCGA database were analyzed, and ARNTL2 expression in LUAD tissues was significantly higher than in normal adjacent tissues (\( P < 0.001 \)). A similar conclusion was reached after combining GTEx data (Fig. 1B). Furthermore, we divided LUAD patients in TCGA database into two equal groups based on ARNTL2 expression. Significant differences in baseline characteristics were found between the two groups, as shown in Table 1 and Supplementary Fig. 1. Patients in the high-ARNTL2 group were more likely to be male and had more advanced N and AJCC stages.

Survival analysis

The TCGA database was used to examine the prognosis of LUAD patients. We discovered that high ARNTL2 expression predicted poor survival, as shown in Fig. 1C: OS: HR = 1.53 (95% CI: 1.14–2.05), \( P = 0.005 \); DFS: HR = 1.35 (95% CI: 1.03–1.78), \( P = 0.031 \). These patients were also subjected to univariate and multivariate analyses to validate survival factors. ARNTL2 (\( p < 0.001 \)), T stage (\( p < 0.001 \)), and N stage (\( p < 0.001 \)) were found to be statistically significant predictors of tumor-specific survival in univariate analyses. ARNTL2 (\( p < 0.001 \)), T stage (\( p < 0.005 \)), and N stage (\( p = 0.001 \)) remained independent prognostic predictors for LUAD patients in multivariate analysis. The forest plot below depicts the relationships between survival outcomes and parameters (Fig. 1D).

TMB and mRNAsi analyses

Tumor mutation burden (TMB) has gotten a lot of attention as one of the immunotherapy biomarkers in recent years [28]. Using the TCGA database, we investigated the relationship between ARNTL2 gene expression and TMB in LUAD patients. A positive correlation was discovered between ARNTL2 and TMB (Fig. 1E); thus, the expression level of ARNTL2 may serve as a reference value for immunotherapy.

It has been reported that mRNAsi is a reliable method for determining the degree of tumor differentiation [29]. The higher mRNAsi values are associated with a higher degree of tumor dedifferentiation and a worse prognosis. As depicted in Fig. 1F, the mRNAsi in LUAD tumor specimens was significantly higher than in normal lung tissues. Furthermore, compared to the low ARNTL2 expression group, the mRNAsi in the high ARNTL2 group was significantly higher, which may be linked to the poor prognosis of patients in the high ARNTL2 group.

Somatic mutations

According to previous research, the number of somatic mutations may be related to tumor development [30]. Fig. 1G depicts the distribution of somatic mutations in the high-ARNTL2 and low-ARNTL2 groups, indicating that TP53 had a high mutation rate in the high-ARNTL2 group (55% vs. 41%, \( P \)-value < 0.01). In contrast, some significant genes, such as RELN (11% in the low subgroup and 18% in the high subgroup), had a lower mutation rate in the high ARNTL2 group.

DEG, GO, KEGG analyses, and single-cell analysis

The prognosis of lung adenocarcinoma patients with varying levels of ARNTL2 expression could be linked to differentially expressed genes (DEGs). So, we analyzed DEG expression to derive a landscape of biological differences (Fig. 2A), and 185 different significant different genes were discovered, including 114 upregulated genes in the high-ARNTL2 group, such as SLC2A1, CD109, and ADGRF4 (adjusted \( P < 0.01 \)), and 71 downregulated genes, such as C16orf89, IRX5, and IRX3 (Supplementary Table 2). The functional enrichment of GO and KEGG in the 185 DEGs was investigated further. Following pathways were linked to the high-ARNTL2 expression: cell cycle, apoptosis, the IL-17 signaling pathway, and the p53 signaling pathway (Fig. 2B). These results were consistent with previous studies [30], which revealed that ARNTL2 is enriched for gene sets associated with LUAD pathogenesis, such as the ‘p53 signaling pathway’ and cell cycle. Furthermore, we used the CancerSEA database to perform a single-cell analysis of lung adenocarcinoma cells, revealing that ARNTL2 gene expression was highly correlated with epithelial-to-mesenchymal transition (EMT), angiogenesis, and metastasis (Fig. 2C, Supplementary Table 3).

IHC and immunofluorescence: ARNTL2 is highly expressed in lung adenocarcinoma tissues and is associated with poor prognosis

The results of immunohistochemistry confirmed the increased expression of ARNTL2 in lung adenocarcinoma. ARNTL2 overexpression was observed in resected tumor tissues compared to adjacent normal tissues, as manifested in Fig. 3A. Immunofluorescence analyses yielded similar results (Fig. 3B). Furthermore, 120 patients from our institution were divided into high and low groups according to the expression of ARNTL2. Patients with high expression of ARNTL2 had significantly poorer OS (\( P \)-value = 0.014, Fig. 3C) and PFS (\( P \)-value = 0.005, Fig. 3C).

ARNTL2 knockdown dramatically inhibits proliferation, invasion, and migration of LUAD cells

To investigate the effect of ARNTL2 on the proliferation and migration of lung adenocarcinoma cells, we created a cell model of ARNTL2 downregulation in A549 and H1299 cells by stably transducing them with ARNTL2 shRNA-expressing lentiviruses. Following that, we used a Western blot to determine the level of ARNTL2 expression after transfection. The knockdown efficiency of shRNA at the protein level was confirmed (Fig. 3D). In A549 and H1299 cell lines, ARNTL2 knockdown was very effective. The sh-ARNTL1 and sh-ARNTL2, which exhibited evident knockdown efficacy, were selected for follow-up experiments. Similarly, overexpression of ARNTL2 in A545 and H1299 cells were verified by Western blot analyses (Fig. 3E).

Subsequently, CCK8, clone formation, wound healing, and transwell assays were used to investigate the role of ARNTL2 in tumor proliferation and migration. CCK8 assays revealed that when ARNTL2 was downregulated compared to control cells, the proliferation ability of both A549 and H1299 cells was reduced (Fig. 3F). Clone formation was used to further demonstrate the effect of ARNTL2 on tumor-forming capability in lung adenocarcinoma cells: ARNTL2-knockdown cells formed significantly fewer colonies than control cells (Fig. 3G).
Fig. 1. A ARNTL2 expression levels in different tumor types from the TCGA database were determined by TIMER. B Differences of ARNTL2 expression in lung adenocarcinoma tissues and normal tissues. C Survival curves comparing the high and low expression of ARNTL2 in lung adenocarcinoma. D Univariate and multivariate analysis of overall survival in LUAD patients from TCGA database. E Scatterplots of correlations between ARNTL2 expression and TMB. F The comparison of mRNAsi in high, low ARNTL2 expression groups of lung adenocarcinoma tissues versus normal tissues. G Differential mutations and their distributions in the high and low-ARNTL2 expression groups.
indicating that ARNTL2 is required for lung adenocarcinoma cell oncogenicity.

The effect of ARNTL2 knockdown on lung adenocarcinoma cell migration and invasion was also confirmed. In a wound-healing assay, ARNTL2-knockdown significantly reduced the migration of A549 and H1299 cells; shControl cells almost recovered the wound within 48 h, whereas ARNTL2-knockdown significantly reduced the migration of A549 and H1299 cell lines. In a transwell migration and invasion assay, ARNTL2-knockdown significantly reduced the migration of A549 and H1299 cell lines. The results showed that the expression levels of NFE2L2, SLC7A11 and CSD1 were significantly decreased after knockdown of ARNTL2, conversely the expression of GLS2 was increased. The opposite result was obtained after overexpressing ARNTL2 (Fig. 4E).

ARNTL2 expression appears to affect drug sensitivity in lung adenocarcinoma, with IC50s for Cisplatin being significantly lower in the high-ARNTL2 group (Fig. 4F). We knocked down ARNTL2 in A549 and H1299 cell lines and examined the cell viability after 48 h exposure to Cisplatin with gradient doses. The results revealed that knockdown of ARNTL2 significantly desensitized both A549 and H1299 cell lines to Cisplatin. Opposite results were obtained upon ARNTL2 overexpression (Fig. 4G).

**ARNTL2 overexpression promotes proliferation, invasion, and migration of LUAD cells**

As shown in Fig. 5, the overexpression of ARNTL2 in both A549 and H1299 cell lines generated an opposite effect compared with ARNTL2 knockdown, leading to increased tumor cell proliferation, tumorigenesis, and migration as well as invasion (Fig. 5A, B, and C).

**ARNTL2 promotes EMT and Tumor formation of LUAD cells**

WB analysis in A549 and H1299 cells confirmed the positive relationship between ARNTL2 and EMT, as shown in Fig. 5D. ARNTL2 knockdown reduced the expression of Snail, Fibronectin, vimentin, and N-cadherin in both cell lines while increasing the expression of E-cadherin and β-catenin. Conversely, the opposite result was found in ARNTL2-overexpressing cells, which indicates Overexpression of ARNTL2 promoted EMT while knockdown of ARNTL2 attenuated EMT.

Tumor formation assays were also performed to determine whether ARNTL2 affects tumor growth in vivo by subcutaneously injecting ARNTL2 knockdown A549 cells or control cells into the flanks of nude mice. The results showed that ARNTL2 shRNA knockdown significantly inhibited tumor growth in the mouse model: ARNTL2 knockdown A549 cells and examined the cell viability after 48 h exposure to Cisplatin with gradient doses. The results revealed that knockdown of ARNTL2 significantly desensitized both A549 and H1299 cell lines to Cisplatin. Opposite results were obtained upon ARNTL2 overexpression (Fig. 4G).

**Discussion**

Many studies have recently shown that circadian rhythm disorders may be linked to the development of various cancers in humans [6,33]. Circadian rhythm disruption can alter the expression of several genes, resulting in dysregulated cell proliferation and subsequent tumorigenesis. There is bidirectional coupling between cell cycle and rhythm genes; several cell cycle genes, including c-Myc, Wee1, cyclin D, and p21, have been reported to be affected by the biological clock [34]. For example, BMAL1, one of the core genes of biological rhythm, can indirectly regulate p21 transcription by regulating REV-ERBα and RORα/γ. CLOCK/BMAL1 also controls the rhythmic expression of WEHI1 kinase, which inhibits the G2/M transition by phosphorylating CDK1 [35]. Furthermore, the circadian clock strongly influences DNA repair, particularly nucleotide excision repair. One of the repair factors, XPA, is regulated by cryptochrome, a common core circadian clock protein

| Table 1 | LUAD patient characteristics according to ARNTL2 expression. |
|---------|-------------------------------------------------------------|
| Total evaluated | high-group | low-group | p value |
| Age(years) | 257 | 256 | 0.931 |
| Mean ± SD | 65.3 ± 10.5 | 65.3 ± 9.8 | 0.046 |
| Sex | | | | |
| male | 130 | 107 | 0.046 |
| female | 127 | 149 | 0.217 |
| Smoking history | | | | |
| No | 34 | 40 | 0.89 |
| Yes | 213 | 212 | 0.001 |
| Unknown | 10 | 4 | 0.001 |
| Pathologic T stage | | | | |
| T1 | 72 | 96 | |
| T2 | 142 | 134 | |
| T3 | 30 | 17 | |
| T4 | 11 | 8 | |
| Unknown | 2 | 1 | |
| Pathologic N stage | | | | |
| N0 | 142 | 188 | 0.099 |
| N1 | 60 | 35 | |
| N2 | 47 | 27 | |
| N3 | 1 | 1 | |
| Unknown | 7 | 5 | |
| Pathologic M stage | | | | |
| M0 | 178 | 166 | |
| M1 | 16 | 9 | |
| Unknown | 63 | 81 | 0.781 |
| A/E 8th stage | | | | |
| 1 | 115 | 159 | |
| 2 | 69 | 52 | |
| 3 | 53 | 31 | |
| 4 | 17 | 9 | |
| Unknown | 3 | 5 | |
| Radiation | | | | |
| Yes | 7 | 6 | 0.350 |
| No/Unknown | 250 | 250 | |
| Chemotherapy | | | | |
| Yes | 82 | 72 | |
| No/Unknown | 175 | 184 | |

SLC7A11, CSD1, and other critical genes associated with ferroptosis were found to be overexpressed in the high-ARNTL2 group (Fig. 4C), implying a possible link between ARNTL2 and ferroptosis in lung adenocarcinoma. We knocked down ARNTL2 in A549 and H1299 cell lines and examined the cell viability after 6 h and 24 h exposure to RSL3 with gradient doses, respectively. The results revealed that following ARNTL2 knockdown, LUAD cells became more sensitive to RSL3 treatment (Fig. 4D). In contrast, the opposite results were observed in ARNTL2-overexpressing A549 and H1299 cell lines. To further verify the effect of ARNTL2 on ferroptosis, we examined the expression levels of several key components involved in ferroptosis pathways after overexpression of ARNTL2 in A549 and H1299 cell lines. The results showed that the expression levels of NFE2L2, SLC7A11 and CSD1 were significantly decreased after knockdown of ARNTL2, conversely the expression of GLS2 was increased. The opposite result was obtained after overexpressing ARNTL2 (Fig. 4E).
Another core clock factor, Per1, controls DNA damage-induced apoptosis by interacting with the checkpoint proteins ATM and Chk2 [37]. ARNTL2, also known as BMAL2, is a paralog of BMAL1 and can dimerize with Clock and induce E-box dependent trans-activation, thereby regulating biological rhythms [36] and inhibiting periodic genes (Per1, Per2) and cryptochrome genes (Cry1 and Cry2) [38]. ARNTL2 has recently been linked to an oncogenic role in various human cancers. However, there has been little research into the role and mechanism of ARNTL2 in lung adenocarcinoma. We discovered and validated the role of ARNTL2 in the development, progression, and treatment of LUAD by combining TCGA and GTEx data. ARNTL2 significantly impacts patients’ clinical features and outcomes with lung adenocarcinoma. The expression of ARNTL2 was higher in LUAD, as demonstrated in our study. ARNTL2 has also been found to be upregulated in BLCA, BRCA, COAD, and READ [39]. We discovered that high ARNTL2 expression was associated with lymph node metastasis in patients with lung adenocarcinoma; similar findings...
have been reported in colorectal cancer [40]. Furthermore, patients with LUAD with high-ARNTL2 expression had a worse prognosis; thus, the high-ARNTL2 expression could be an independent predictor of prognosis in lung adenocarcinoma.

We confirmed that ARNTL2 promoted LUAD cell proliferation, migration, and invasion using in vivo and in vitro functional experiments. The mechanism is most likely multifactorial. Our single-cell analysis revealed that ARNTL2 expression was positively associated with EMT and metastasis, confirmed by a Western blot of important EMT marker proteins. A previous study demonstrated that downregulation of ARNTL2 in colon carcinoma could suppress tumor cell proliferation and migration via SMOC2-EMT by inactivation of the PI3K/AKT signaling pathway [10]. Brady et al. also discovered that ARNTL2 works with Clock, another circadian gene, to drive the expression of a complex pro-metastatic secretome that promotes lung adenocarcinoma cell metastasis and growth [41]. Our results also demonstrated, ARNTL2 can inhibit ferroptosis of LUAD cells. And LUAD cells with varied ARNTL2 levels (NC, ARNTL2-knockdown, and ARNTL2-overexpression) showed significant expressing difference in some key ferroptosis-related genes, such as NFE2L2, SLC7A11, and CISD1.

Immunotherapy is gaining popularity in lung cancer treatment. Tumor immune infiltration, and immune checkpoints were all associated with ARNTL2 expression levels in lung adenocarcinoma, indicating ARNTL2 may affect the tumor immune microenvironment of lung adenocarcinoma. For example, ARNTL2 expression was strongly positively correlated with CD8+ T cells but strongly negatively correlated with Natural killer (NK) cells. We also discovered significant differences in the expression of immune checkpoints between patients in the high-
ARNTL2 and low-ARNTL2 groups. More research into the relationship between ARNTL2 and immune checkpoints could be beneficial and necessary.

This study had some limitations as well. Although a large cohort of TCGA and GTEx databases was used to develop and validate the role of ARNTL2 in lung adenocarcinoma, but due to the retrospective nature of our study, design selection bias could not be avoided. Furthermore, this study did not conduct additional in-depth experiments to investigate the mechanism of ARNTL2 promoting tumor development and progression in lung adenocarcinoma, which is the direction of our future research.

**Conclusion**

ARNTL2 is highly expressed in lung adenocarcinoma and was found to be an independent predictor of a poor prognosis in patients with lung adenocarcinoma. Functional experiments showed that ARNTL2 promoted the cell proliferation, migration, and invasion in vitro as well as tumor growth in vivo. Notably, we demonstrated for the first time that ARNTL2 promotes EMT and was linked to ferroptosis and cell sensitivity to cisplatin in lung adenocarcinoma cells.
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Ethics approval and consent to participate

All methods were carried out in accordance with relevant guidelines and regulations. All participants signed informed consent according to the ethical requirements in the Declaration of Helsinki. Ethics approved by the ethical committees of Zhongshan Hospital (B2018–137R and Y2020–529).

Consent for publication

Not applicable.

Availability of data and materials

The datasets used and analyzed during the current study are available from the corresponding author on reasonable request.

Disclosures

The authors declare that there is no conflict of interest.

Fig. 5. A-B The effects of ARNTL2 overexpressing on cell proliferation (A), and Comparison of colony formation (B) in A549 and H1299. C The effects of ARNTL2 overexpressing on cell invasion and migration in A549 and H1299 cells. D Expressions of EMT-related markers in A549 and H1299 cells with ARNTL2 knockdown and overexpression by WB assays: indicating a positive correlation between ARNTL2 expression and EMT. E The effects of ARNTL2 knockdown on tumor growth in mouse xenograft model: ARNTL2 knockdown markedly suppressed xenograft tumor growth of A549 cells.
CRediT authorship contribution statement

Huan Zhang: Methodology, Validation, Writing – original draft.
Guangyao Shan: Visualization, Writing – review & editing, Validation.
Xing Jin: Methodology, Software. Xiangyang Yu: Validation. GuoShu Bi: Methodology, Visualization. Xiangyang Feng: Visualization. Hao Wang: Visualization, Writing – review & editing. Miao Lin: Supervision. Cheng Zhan: Methodology, Visualization. Qun Wang: Data curation, Funding acquisition. Ming Li: Conceptualization, Methodology, Writing – review & editing, Resources.

Declaration of Competing Interest

There are no conflicts of interest to declare.

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Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.tranon.2022.101562.

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