

# Specific association of *MTHFD1* expressions with small cell lung cancer development and chemoradiotherapy outcome

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## ABSTRACT

**الأهداف:** تحديد المؤشرات الحيوية التي يمكن أن تميز سرطان الرئة ذو الخلايا الصغيرة (SCLC) عن غير (NSCLC)، واستكشاف ارتباطها بتشخيص SCLC تحت العلاج الكيميائي.

**المنهجية:** استخدمنا مجموعة البيانات GSE40275 لتحديد الأهداف المحتملة في SCLC. كان هناك 196 مريضاً بسرطان الرئة (LC) في المجموعة الأولى من هذه الدراسة. تم تحديد مستويات *MTHFD1* في الأنسجة بواسطة مقايسة الكيمياء المناعية في الفوج 1. ادرجنا كذلك مرضى سرطان الرئة الذين خضعوا جميعاً للعلاج الكيميائي المحلي (CRT) في الفوج 2، وتم تحديد ارتباط مستويات *MTHFD1* مع نتائج علاج CRT في الفوج 2. أجرينا تجارب الخلايا لتحديد وظيفة *MTHFD1* على حساسية الأشعة لخلايا SCLC و NSCLC.

**النتائج:** تمت زيادة مستويات *MTHFD1* في أنسجة LC، ويمكن أن تميز SCLC عن كل من سرطان الخلايا الحشرية الرئوية (LUSC) وسرطان الرئة الغدي (LUAD). كان لدى مرضى سرطان الرئة ذو الخلايا الصغيرة الذين يعانون من النمط الظاهري العالي *MTHFD1* تشخيصاً أسوأ بعد علاج CRT، في حين لم يتم العثور على ارتباط كبير بين مستويات *MTHFD1* والتشخيص في مجموعة LUSC و LUAD. أظهرت تجارب الخلايا أن الإفراط في التعبير عن *MTHFD1* يزيد من المقاومة الإشعاعية في كل من SCLC و NSCLC في المختبر.

**الخلاصة:** قد تكون تعبيرات *MTHFD1* علامة بيولوجية تشخيصية جديدة على وجه التحديد لـ SCLC ونتائج علاج CRT.

**Objectives:** To identify biomarkers that can discriminated small cell lung cancer (SCLC) from non-SCLC (NSCLC), and explore their association with the prognosis of SCLC under chemoradiotherapy.

**Methods:** The GSE40275 dataset was used to identify potential targets in SCLC. There were 196 patients of lung cancer (LC) in cohort 1 of this study. *MTHFD1* levels in tissues were determined by immunohistochemistry assay in cohort 1. Lung cancer patients who were all underwent local chemoradiotherapy (CRT) were included in cohort 2, and the association of *MTHFD1* levels with CRT treatment outcome were determined in cohort 2. Cell experiments were used to determine the function

of *MTHFD1* on the radio-sensitivity of SCLC and NSCLC cells.

**Results:** The *MTHFD1* levels in LC tissues were increased, and could discriminate SCLC from both lung squamous cell carcinoma (LUSC) and lung adenocarcinoma (LUAD). Small cell lung cancer patients with *MTHFD1* high phenotype had a poorer prognosis after CRT treatment, whereas no significant correlation was found between *MTHFD1* levels and prognosis in LUSC and LUAD group. Cell experiments demonstrated that overexpression of *MTHFD1* increases radio-resistance in both SCLC and NSCLC in vitro.

**Conclusion:** *MTHFD1* expressions might be a novel specifically prognostic biomarker for SCLC and the CRT treatment outcome.

**Keywords:** biomarker, SCLC, LUSC, LUAD, prognosis

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Lung cancer (LC) is not a single entity but rather a complex disease with distinct subtypes, primarily categorized into 2 major groups: non-small cell lung cancer (NSCLC) and small cell lung cancer (SCLC).<sup>1,2</sup> Non-small cell lung cancer mainly includes lung squamous cell carcinoma (LUSC) and lung adenocarcinoma (LUAD). Generally, advanced age and positive smoking status, male gender, SCLC pattern, and advanced lung cancer stage (III or IV) are associated with poor prognosis.<sup>3,4</sup>

There are huge differences in molecular phenotypes between SCLC and NSCLC, and many studies have focused on identifying SCLC-specific genes and their biological functions to uncover the regulatory mechanisms and potential biomarkers for SCLC.<sup>5</sup> Currently, the GSE40275 dataset, which is the RNA sequencing (RNA-seq) data determining the genes profiles of different subtypes of LC, has been uploaded to the GEO dataset. Through bioinformatics analysis of this dataset, we identified a series of SCLC related genes, including *MTHFD1*.

*MTHFD1* is a gene that encodes a trifunctional enzyme involved in folate metabolism.<sup>6</sup> Human *MTHFD1* expressions and polymorphisms are reported to be closely associated with various malignancy.<sup>7-9</sup>

For SCLC and advanced NSCLC, a combined approach of radiotherapy and chemotherapy (chemoradiotherapy [CRT]) is the primary choice. However, there is a significant variation in the effectiveness of the CRT treatment. Here, we plan to investigate the relationship between *MTHFD1* expression and the develop of different subtypes of LC, and their CRT treatment outcome.

**Methods.** This study is a prospective study, carried out from January 2020 to October 2023. The Eighth Medical Center of the Chinese PLA General Hospital, Beijing and Qingdao Municipal Hospital, Qingdao, China, were included in this study. The inclusion criteria were: I) histopathologically diagnosed LUAD, LUSC or SCLC; II) age 18-75 years; and III) complete clinical data. The exclusion criteria were: I) complicated with other malignant tumors; II) liver or kidney dysfunction; and III) pregnant or lactating women. A total of 196 LC patients, including 70 LUAD, 56 LUSC, and 70 SCLC cases were recruited as cohort 1, and their pathological tissues were used for subsequent detection.

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A total of 33 patients with inflammatory pseudotumor were included and their normal lung epithelial tissue samples were used as controls (HCs). Next, we selected patients who underwent radiotherapy and chemotherapy (excluding those who did not undergo radiotherapy and chemotherapy) in cohort1 to form cohort 2 (cohort 2 includes: 52 SCLCs, 30 LUADs, and 31 LUSCs). The study was carried out according to the Helsinki guidelines, with the approval from The ethics committee of the Eighth Medical Center of the Chinese PLA General Hospital, Beijing, China (approval ID: 3092023323152361). All participants provided their written consent.

Lung cancer tissues were homogenized, and then TRIzol™ LS Reagent (Thermo Fisher Scientific, Waltham, MA) were used to extract the total RNA from these tissues. Subsequently, the qRT-PCR detection was carried out by using PrimeScript RT-PCR kit (TaKaRa, Osaka, Japan) and SYBR Green PCR Kit (TaKaRa, Osaka, Japan) and *MTHFD1* mRNA levels were determined by using the following primers:

Forward: 5'-GTTGAAGGAGCAAGTACCTGG-3';

Reverse: 5'-GGTAGCTGCACTAAGAACCCA-3';

Immunohistochemical (IHC) testing was carried out according to previous reports.<sup>7</sup> The LC tissues or normal lung tissues samples were uniformly cut into 4 μm paraffin sections. Two experienced pathologists will independently interpret the IHC results. Scoring of the IHC will be based on the integrated optical density (IOD) measurement.<sup>7</sup> Then, the expression of *MTHFD1* will be uniformly normalized.

The human SCLC cell line H69, NSCLC cell line A549, and normal lung epithelial cell line BEAS-2B were purchased from the American Type Culture Collection (ATCC, USA). All cells were incubated in a humidified atmosphere containing 5% CO<sub>2</sub> at 37°C with DMEM medium supplemented with 10% fetal bovine serum (GIBCO, USA).

For the radio-resistance assay, cells were instantly transfected with *MTHFD1* over-expressed plasmid or control plasmid by using Lipofectamine2000 for 24 hours before being exposed to different doses of irradiation. After 48 hours post-irradiation, MTT assay was carried out using the commercialized MTT kit (Sigma, USA) in accordance with the instruction manual.

The GSE40275 dataset, which describes the transcripts expressions in SCLC, LUSC, and LUAD samples, was obtained from GEO and were analyzed by GEO2R tool, Venn diagram and Metascape.<sup>10,11</sup>

**Statistical analysis.** The differences between groups were determined by using The Statistical Package for the

Social Sciences, version 26 (IBM Corp, Armonk, NY, USA). Receiver operating characteristic (ROC) curves were used to evaluate the performance of diagnostic tests. Kaplan-Meier was used to create the overall survival (OS) curve.

**Results.** The GSE40275 dataset mapped the global transcriptome sequencing results of tissue samples derived from SCLC and NSCLC (mainly including LUSC and LUAD).<sup>10</sup> With an adj. *p*-value of <0.05, we identified 16323 differentially expressed targets between the SCLC and LUSC (6420 up-regulated and 9908 down-regulated), and 12004 differentially expressed targets between the SCLC and LUSC (5133 up-regulated and 6871 down-regulated; **Figure 1A&B**). We found that there were only 330 differentially expressed genes between LUSC and LUAD. Above results indicated significant genetic differences both between SCLC and LUSC and between SCLC and LUAD, while molecular phenotypes of LUAD and LUSC are relatively similar.

The Venn diagram showed 11004 overlapping targets between the 2 comparisons (4722 up-regulated and 6282 down-regulated; **Figure 1C&D**). The main biological processes related to the overlapping targets were mapped using GO analysis (**Figure 1C**), and it indicated that these targets were highly involved in the mitochondrial translation, DNA or RNA methylation and methionine biosynthetic process (**Figure 1E**). Therefore, we focused on the genes that were involved in both mitochondrial translation and DNA or RNA methylation.

We next used a small-scale samples to quickly verify the bioinformatics analysis results above. We included 18 paired tissues (tumor tissues and matched adjacent tissues) from 6 cases of SCLC, 6 cases of LUSC and 6 cases of LUAD, who were randomly selected from participants in cohort 1. Total RNAs were extracted from all the tissues above, and extract their total RNAs. We focused on *DNMT1*, *MAT2A*, *HDAC5*, *METTL3*, *MECP2*, and *MTHFD1* because they are all involved in both mitochondrial translation and DNA or RNA methylation, and ranked high in differential expression analysis. We found that *MTHFD1*, *DNMT1*, and *METTL3* were higher in SCLC tissues than those in LUSC and LUAD tissues, and *MTHFD1* has the greatest expression difference (**Figure 2A&B**). Therefore, we next explore the role of *MTHFD1* using a large sample size.

We included 196 LC patients in cohort 1 to delve into the expression characteristics of *MTHFD1*. Detailed demographics for all the subjects are listed in

**Table 1.** We also enrolled 33 gender and age matched HCs. A total of 56 patients were LUSC, 70 were LUAD, and 70 patients were SCLC.

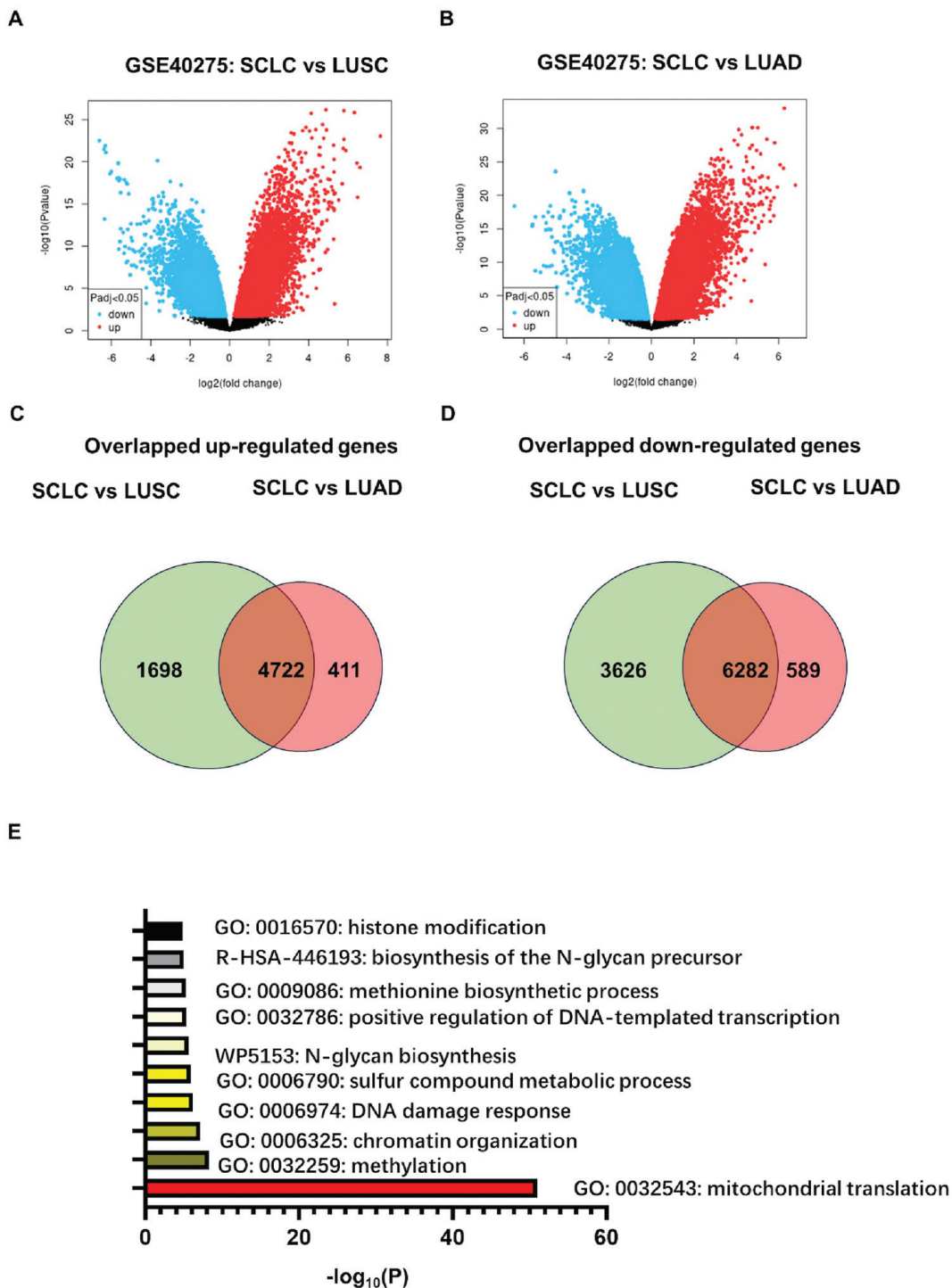
Immunohistochemical results show that the *MTHFD1* increased in cancer tissues compared with the adjacent tissues, regardless of their cancer tissue types, including SCLC, LUSC, and LUAD (**Figure 2C**). Meanwhile, we confirmed that the expression of *MTHFD1* in all the LC types were significantly higher than that in HCs, and it was particularly upregulated in SCLC, compared to LUSC and LUAD (**Figure 2D**).

Receiver operating characteristic curves were next constructed. The *MTHFD1* levels discriminated LUSC, LUAD, and SCLC patients from healthy control patients (**Figure 2E-G**). And, *MTHFD1* could discriminate SCLC from LUSC and LUAD patients (**Figure 2H&I**). These results indicated that *MTHFD1* might act as a biomarker for the diagnosis of both NSCLC and SCLC, and could discriminate SCLC from NSCLC.

Next, SCLC, LUSC, and LUAD patients were allocated into *MTHFD1* high group and low group. Among 70 SCLC, 56 LUSC and 70 LUAD patients, 31 SCLC patients, 22 LUSC patients and 25 LUAD patients were *MTHFD1* high phenotype and others were *MTHFD1* low cases. For patients with SCLC, *MTHFD1* high phenotype was associated with T1 stage and distant metastasis. However, *MTHFD1* high phenotype was not associated with any indicators in LUSC and LUAD patients (**Table 2**).

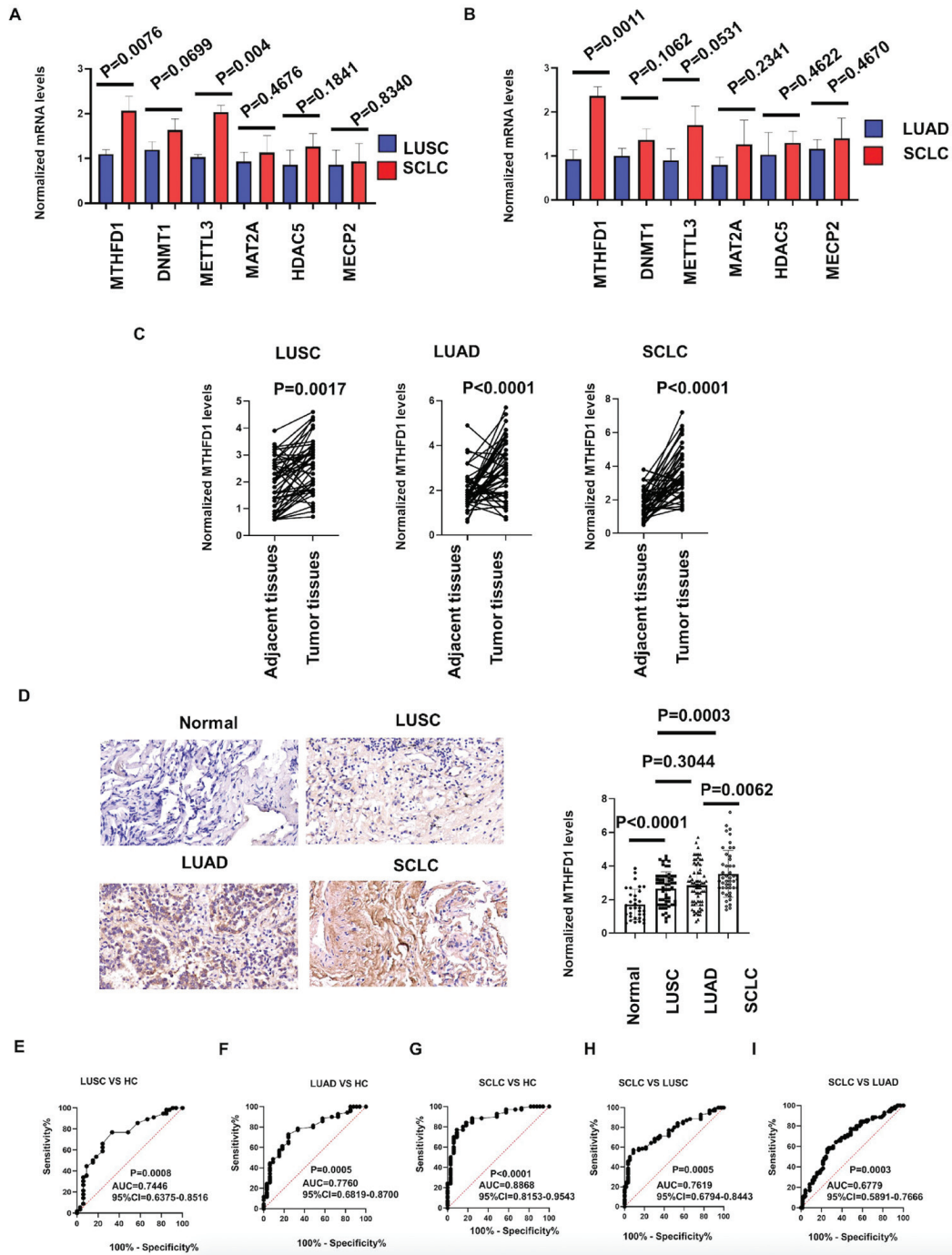
Next, we explored the relationship between *MTHFD1* levels and LC prognosis after CRT treatment. We used cohort 2, which were all from cohort 1, except for excluding the non-CRT treated in cohort 1. The results showed a visible worse OS rate in the SCLC group, compared with both LUSC group and LUAD group. And, for the patients with SCLC, *MTHFD1* high phenotype was associated with a poor 2-year OS curve (**Figure 3A**), whereas for the patients with LUSC and LUAD, the 2-year OS curve was not significantly associated with the *MTHFD1* levels (**Figure 3B&C**). Collectively, these results indicated that *MTHFD1* levels were specifically associated with prognosis of SCLC.

The *MTHFD1* expression was subsequently analyzed by ELISA in cultured cell lines of SCLC (H69), NSCLC (A549) and normal lung epithelial cells (BEAS-2B). We found that it was increased in cancer cell line, and were expressed higher in H69 than A549, which is in accordance with clinical findings (**Figure 4A**). To determine whether *MTHFD1* regulated



**Figure 1** - Differential transcripts between small cell lung cancer (SCLC) and lung squamous cell carcinoma (LUSC), and between SCLC and lung adenocarcinoma (LUAD), and the overlapped targets based on the GSE40275 dataset. The volcano maps showing the differential transcripts between A) SCLC and LUSC; and B) between SCLC and LUAD, based on the GSE40275 dataset. The GEO2R tool was used to identify differentially expressed genes. The Venn diagram showing the overlapped C) up-regulated transcripts; and D) down-regulated transcripts among the differential transcripts described in A & B. E) GO enrichment of the overlapped differentially expressed transcripts described in C & D. The plot was calculated and drawn by Metascape Gene List Analysis. LUSC: lung squamous cell carcinoma, LUAD: lung adenocarcinoma, SCLC: small cell lung cancer





**Figure 2 -** *MTHFD1* was higher expressed in small cell lung cancer (SCLC) tissues than lung squamous cell carcinoma (LUSC) and lung adenocarcinoma (LUAD) tissues. Comparing the mRNA levels of *DNMT1*, *MAT2A*, *HDAC5*, *METTL3*, *MECP2*, and *MTHFD1* between: A) SCLC tissues (n=6) and LUSC tissues (n=6); and B) between SCLC tissues (n=6) and LUAD tissues (n=6) in a small-scale samples. C) Comparing the *MTHFD1* levels between tumor tissues and matched adjacent tissues in LUSC (left, n=56), LUAD (middle, n=70), and SCLC (right, n=70) cases in a large-scale samples. D) Comparing the *MTHFD1* levels in tissues from healthy controls (HCs, n=33), LUSC (n=56), LUAD (n=70), and SCLC (n=70) cases in a large-scale samples. E-I) The receiver operating characteristic curve for the *MTHFD1* levels in distinguishing: E) LUSCs and HCs; F) LUADs and HCs; G) SCLCs and HCs; H) SCLC and LUSCs; and I) SCLC and LUADs. Left: representative stained images among these 4 groups, right: bar chart comparison of differences among these 4 groups, HCs: healthy controls, LUSC: lung squamous cell carcinoma, LUAD: lung adenocarcinoma, SCLC: small cell lung cancer

the radio-sensitivity of SCLC cells, the viability and cell apoptosis of H69 and A549 after exposure to various doses of irradiation was analyzed. Cell viability was inhibited (Figure 4B&C), and cell apoptosis was induced (Figure 4D&E), by irradiation in a dose-dependent manner, and overexpression of *MTHFD1* led to a markedly lower level of radiation-induced cell death and cell apoptosis, compared with the control cells, in

both H69 and A549 cells (Figure 4B-E). These results indicated that *MTHFD1* increases radio-resistance in both SCLC and NSCLC cells.

**Discussion.** Biomarkers are needed that enable earlier detection and monitoring of LC.<sup>12</sup> The RNA-seq data for analyzing LC samples can accelerate the development process of biomarker. For example, based on the integrating single-cell RNA-seq and bulk RNA-seq, Zhang et al<sup>13</sup> identified new therapeutic targets for patients with advanced LUAD. Similarly, Lavanya et al<sup>14</sup> screened a series of potential specific biomarkers for SCLC by mining existing RNA-seq data. The GSE40275 dataset provides RNA-seq data for different tissue subtypes such as SCLC, LUSC, and LUAD, and some groups have made progress in LC research by analyzing this dataset.<sup>10,15,16</sup> The starting point of this study is to screen targets by comparing the RNA-seq data between SCLC and LUSC, as well as between SCLC and LUAD, and taking their intersections, which will increase the probability of discovering SCLC specific biomarkers. After that, we screened many targets highly involved in mitochondrial translation, DNA or RNA methylation and methionine biosynthetic process.

The *MTHFD1* ranked high among the targets, and was confirmed by us to be up-regulated in SCLC, which could discriminate SCLC from LUSC and LUAD. The *MTHFD1* was the primary source of folate-activated one-carbon units in the cytoplasm. Increased evidence has been prompted that *MTHFD1* was responsible for variable tumor. Yu H et al<sup>7</sup> confirmed that overexpression of *MTHFD1* in hepatocellular carcinoma could predict poorer survival and recurrence. In our study, we found

**Table 1 -** Clinical indications for included lung cancer patients.

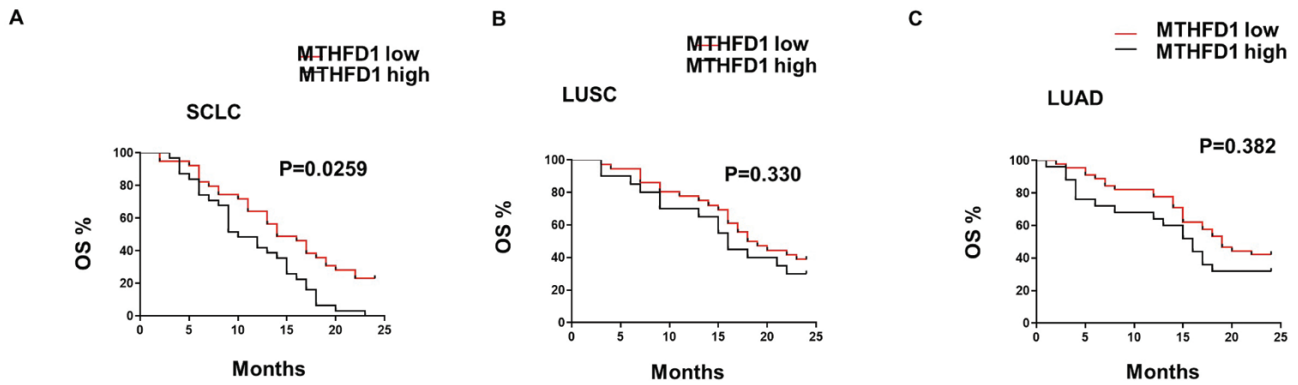
Variables	LC (n=196)	HC (n=33)
<i>Age</i>		
Above 60	121 (61.7)	20 (60.6)
Below 60	75 (38.3)	13 (39.4)
<i>Gender</i>		
Male	139 (70.9)	20 (60.6)
Female	57 (29.1)	13 (39.4)
<i>T stage</i>		
1-2	94 (48.0)	---
3-4	102 (52.0)	---
<i>Metastases</i>		
<i>LM</i>		
Yes	145(74.0)	---
No	51 (26.0)	---
<i>DM</i>		
Yes	96 (48.9)	---
No	100 (51.1)	---
<i>Tumor subtype</i>		
LUAD	70 (35.7)	---
LUSC	56 (28.6)	---
SCLC	70 (35.7)	---

Values are presented as numbers and percentages (%). LC: lung cancer, LUAD: lung adenocarcinoma, LUSC: lung squamous cell carcinoma, SCLC: small cell lung cancer, LM: lymphatic metastasis, DM: distant metastasis

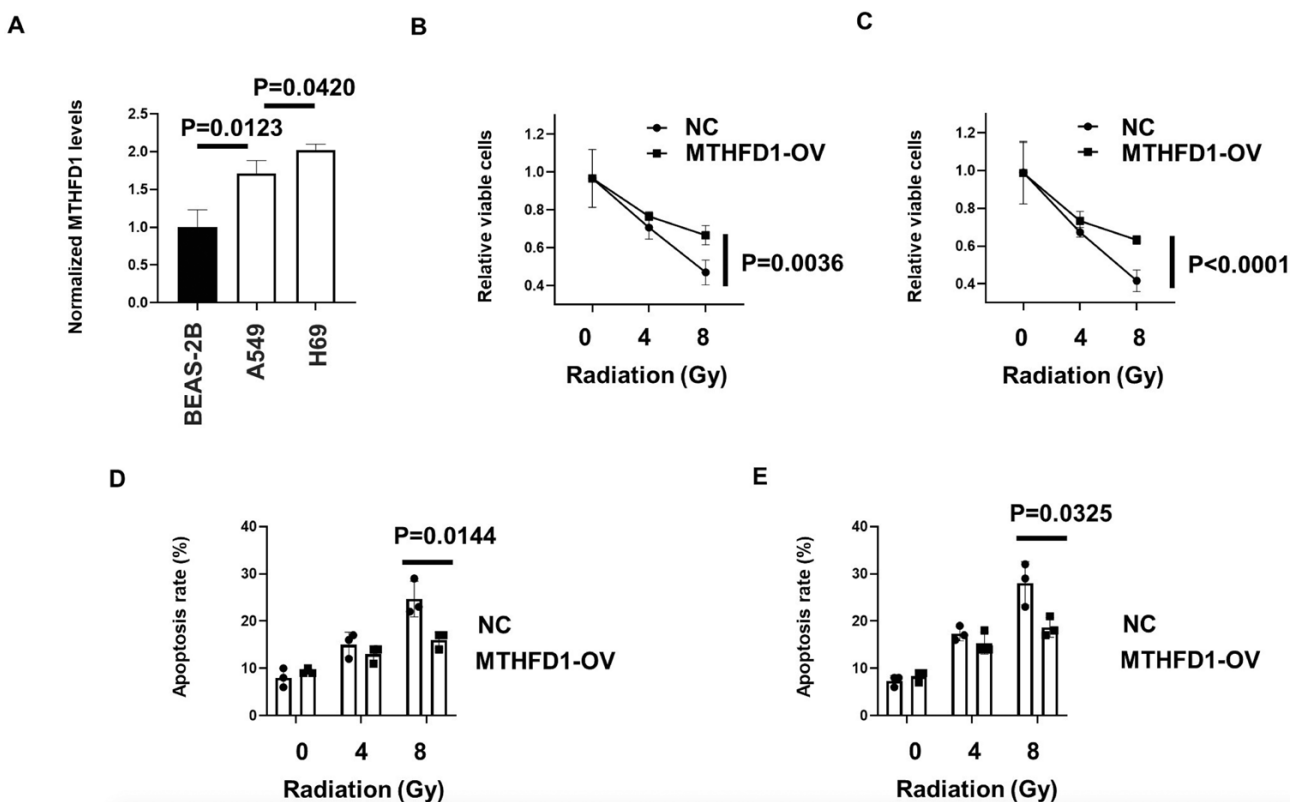
**Table 2 -** Association of *MTHFD1* levels with clinical indications for small cell lung cancer, lung squamous cell carcinoma and lung adenocarcinoma patients.

Variables	SCLC (n=70)			LUSC (n=56)			LUAD (n=70)		
	<i>MTHFD1</i> high (n=31)	<i>MTHFD1</i> low (n=39)	<i>P</i> -values	<i>MTHFD1</i> high (n=22)	<i>MTHFD1</i> low (n=34)	<i>P</i> -values	<i>MTHFD1</i> high (n=25)	<i>MTHFD1</i> low (n=45)	<i>P</i> -values
<i>T stage</i>									
1-2	9 (29.0)	25 (64.1)	0.003	9 (40.9)	20 (50.8)	0.190	8 (32.0)	25 (55.6)	0.058
3-4	22 (71.0)	14 (35.9)		13 (59.1)	14 (41.2)		17 (68.0)	20 (44.4)	
<i>Metastases</i>									
<i>LM</i>									
Yes	25 (80.6)	27 (69.2)	0.278	18 (81.8)	26 (76.5)	0.634	20 (80.0)	29 (64.4)	0.173
No	6 (19.4)	12 (30.8)		4 (18.2)	8 (23.5)		5 (20.0)	16 (35.1)	
<i>DM</i>									
Yes	20 (64.5)	15 (38.5)	0.030	14 (63.6)	13 (38.2)	0.063	15 (60.0)	19 (42.2)	0.154
No	11 (35.5)	24 (61.5)		8 (36.4)	21 (61.8)		10 (40.0)	26 (57.8)	

Values are presented as numbers and percentages (%). LC: lung cancer, LUAD: lung adenocarcinoma, LUSC: lung squamous cell carcinoma, SCLC: small cell lung cancer, LM: lymphatic metastasis, DM: distant metastasis



**Figure 3** - *MTHFD1* levels were specifically associated with prognosis of small cell lung cancer (SCLC) patients after chemoradiotherapy treatment. Kaplan-Meier curves for time to 2-year overall survival of patients with: A) SCLC; B) lung squamous cell carcinoma; and C) lung adenocarcinoma according to *MTHFD1* levels. LUSC: lung squamous cell carcinoma, LUAD: lung adenocarcinoma, SCLC: small cell lung cancer, OS: overall survival



**Figure 4** - Overexpression of *MTHFD1* increases radio-resistance in both small cell lung cancer and non-small cell lung cancer in vitro. A) *MTHFD1* levels were determined in H69, A549, and BEAS-2B cells. B-C) The H69 and A549 cells were instantly transfected with *MTHFD1* over-expressed plasmid for 24 hours, and then the viable cells of B) H69; and C) A549 were detected after treated by different dose irradiation. D-E) The H69 and A549 cells were instantly transfected with *MTHFD1* over-expressed plasmid and control plasmid for 24 hours, and then the apoptosis rate of B) H69; and C) A549 were detected after treated by different dose irradiation.

that high *MTHFD1* levels were specifically associated with poor prognosis of SCLC, but not with LUAD and LUSC, after CRT treatment. In terms of mechanism, *MTHFD1*/folate/one-carbon network is required for

the de novo synthesis of 3 of the 4 DNA bases and the re-methylation of homocysteine to methionine, and therefore exerts its effects in tumor cells.<sup>17,18</sup> It is reported that aberrant DNA methylation, causing gene

silencing, is common in cancers and would induce CRT resistance.<sup>19</sup> And, some studies have reported the significant differences in methylation levels between SCLC and NSCLC.<sup>20</sup> Correspondingly, our clinical data and experimental data showed that the expressions of *MTHFD1* in SCLC cells are significantly higher than that in NSCLC cells, and high expression of *MTHFD1* reduces the sensitivity of radiation to cancer cells.<sup>21</sup> These might be the major reason why *MTHFD1* levels can discriminate SCLC from LUSC and LUAD, and specifically reflect the prognosis of SCLC.

**Study limitations.** this work is just an initial exploratory study, so it has limitations due to a relatively small sample size. We are carrying out a broader study with a more diverse range of samples provided by multiple medical centers to substantiate our findings.

In conclusion, high *MTHFD1* expressions could distinguish SCLC from LUSC and LUAD, and was associated with poor CRT treatment outcome of SCLC. The implication of this study is that it provides a new understanding of the relationship between *MTHFD1* expression and LC progressions, indicating that *MTHFD1* levels might be a specific indicator of SCLC.

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## References

- Bade BC, Dela Cruz CS. Lung cancer 2020: epidemiology, etiology, and prevention. *Clin Chest Med* 2020; 41: 1-24.
- Wu F, Wang L, Zhou C. Lung cancer in China: current and prospect. *Curr Opin Oncol* 2021; 33: 40-46.
- Tseng JS, Chiang CJ, Chen KC, Zheng ZR, Yang TY, Lee WC, et al. Association of smoking with patient characteristics and outcomes in small cell lung carcinoma, 2011-2018. *JAMA Netw Open* 2022; 5: e224830.
- Larsson SC, Burgess S. Appraising the causal role of smoking in multiple diseases: a systematic review and meta-analysis of Mendelian randomization studies. *EBioMedicine* 2022; 82: 104154.
- Baumeister SE, Baurecht H, Nolde M, Alayash Z, Gläser S, Johansson M, et al. Cannabis use, pulmonary function, and lung cancer susceptibility: a Mendelian randomization study. *J Thorac Oncol* 2021; 16: 1127-1135.
- Dekhne AS, Hou Z, Gangjee A, Matherly LH. Therapeutic targeting of mitochondrial one-carbon metabolism in cancer. *Mol Cancer Ther* 2020; 19: 2245-2255.
- Yu H, Wang H, Xu HR, Zhang YC, Yu XB, Wu MC, et al. Overexpression of *MTHFD1* in hepatocellular carcinoma predicts poorer survival and recurrence. *Future Oncol* 2019; 15: 1771-1780.
- Zheng Y, Zhu L, Qin ZY, Guo Y, Wang S, Xue M, et al. Modulation of cellular metabolism by protein crotonylation regulates pancreatic cancer progression. *Cell Rep* 2023; 42: 112666.
- Guan J, Li M, Wang Y, Zhang Y, Que Y, Lu S, et al. *MTHFD1* regulates the NADPH redox homeostasis in MYCN-amplified neuroblastoma. *Cell Death Dis* 2024; 15: 124.
- Zhou Y, Zhou B, Pache L, Chang M, Khodabakhshi AH, Tanaseichuk O, et al. Metascape provides a biologist-oriented resource for the analysis of systems-level datasets. *Nat Commun* 2019; 10: 1523.
- Khaki AA, Tanoomand A, Hajibemani A, Abouhamzeh B. Detection of polymorphisms in *MTHFD1 G1958A* and its possible association with idiopathic male infertility. *Urol J* 2019; 16: 586-591.
- Seijo LM, Peled N, Ajona D, Boeri M, Field JK, Sozzi G, et al. Biomarkers in lung cancer screening: achievements, promises, and challenges. *J Thorac Oncol* 2019; 14: 343-357.
- Chemi F, Pearce SP, Clipson A, Hill SM, Conway AM, Richardson SA, et al. cfDNA methylome profiling for detection and subtyping of small cell lung cancers. *Nat Cancer* 2022; 3: 1260-1270.
- Zhang P, Pei S, Gong Z, Feng Y, Zhang X, Yang F, et al. By integrating single-cell RNA-seq and bulk RNA-seq in sphingolipid metabolism, *CACYBP* was identified as a potential therapeutic target in lung adenocarcinoma. *Front Immunol* 2023; 14: 1115272.
- C L, S P, Kashyap AH, Rahaman A, Niranjana S, Niranjana V. Novel biomarker prediction for lung cancer using random forest classifiers. *Cancer Inform* 2023; 22: 11769351231167992.
- Zhang Y, Chen Q, Huang T, Zhu D, Lu Y. Bioinformatics-based screening of key genes for transformation of tyrosine kinase inhibitor-resistant lung adenocarcinoma to small cell lung cancer. *Front Med (Lausanne)* 2023; 10: 1203461.
- Zhang X, Yang L, Huang B, Yin J, Wei Y. Identification and validation of cyclin A2 and cyclin E2 as potential biomarkers in small cell lung cancer. *Oncol Res Treat* 2023; 46: 246-258.
- Liu M, Xu K, Saaoud F, Shao Y, Zhang R, Lu Y, et al. 29 m<sup>6</sup>A-RNA methylation (epitranscriptomic) regulators are regulated in 41 diseases including atherosclerosis and tumors potentially via ROS regulation - 102 transcriptomic dataset analyses. *J Immunol Res* 2022; 2022: 1433323.
- Meng Q, Lu YX, Wei C, Wang ZX, Lin JF, Liao K, et al. Arginine methylation of *MTHFD1* by *PRMT5* enhances anoikis resistance and cancer metastasis. *Oncogene* 2022; 41: 3912-3924.
- Ponomaryova AA, Schegoleva AA, Gervas PA, Pancova OV, Gerashchenko TS, Zarubin AA, et al. DNA methylome analysis reveals potential alterations contributing to the progression of bronchial hyperplasia. *Mol Biol Rep* 2023; 50: 7941-7947.
- Chen Y, Liu X, Li Y, Quan C, Zheng L, Huang K. Lung cancer therapy targeting histone methylation: opportunities and challenges. *Comput Struct Biotechnol J* 2018; 16: 211-223.