Development of a Prognostic Nomogram for Patients with Lung Adenocarcinoma in the Stages I, II, and III Based on Immune Scores

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Background: Immunotherapy has significantly changed the treatment prospects of non-small cell lung cancer (NSCLC). However, there is no report based on immune score to predict the overall survival (OS) of lung adenocarcinoma (LUAD) in the stages I, II, and III. Therefore, this study aimed to investigate the immune score and the prognosis-related factors of LUAD and construct a nomogram to predict the prognosis.

Methods: A total of 390 cases with lung adenocarcinoma in the stages I, II, and III were included in the study. The clinicopathological characteristics and immune scores of LUAD patients were downloaded from the TCGA database. Cox proportional hazards regression model was used to estimate hazard ratio (HR) and 95% confidence interval (CI). A Nomogram was composed of the Cox model and internally validated using 1000 bootstrap. The concordance index (c-index) and the calibration curves were used to evaluate the model. The decision curve analysis (DCA) was performed to evaluate the clinical practical value of the model.

Results: According to the immune score, the patients were divided into low-, medium-, and high-score groups. This study showed that compared with patients with low and medium immune scores, only patients with high immune scores had significantly improved OS (HR and 95% confidence interval (CI): 0.489 [0.324-0.737]). The C-index for OS prediction was 0.691 (95% CI, 0.646-0.736). The calibration curves for nomogram-predicted probabilities of 3- and 5-year survival have good ability for the calibration and discrimination.

Conclusion: The high immune score was significantly correlated with better OS of patients with LUAD in the stages I, II, and III. Moreover, the nomogram of predicting prognosis may help assess the survival of LUAD patients.

Keywords: lung adenocarcinoma, immune scores, nomograms, prognosis

Background

Updated statistics of lung cancer in China (2019.1) showed that lung cancer had the highest incidence and mortality rate, ranking first among all types of malignant tumors in China. It was estimated that there were 787,000 new lung cancer cases and 631,000 deaths from lung cancer in China in 2015.1 The recently released data of the global cancer burden in 2017 showed that the incidence rate of lung cancer ranked second among all types of cancer, and the mortality rate of lung cancer was the highest in the world.2

The standard treatment of lung cancer is multidisciplinary, including traditional surgery, chemotherapy and radiotherapy, which are still effective methods for the treatment of lung cancer.
In recent years, significant progress has been made in molecular targeted therapy and immunotherapy of lung cancer. These new methods have brought major changes to the treatment of lung cancer and improved the quality of life and survival rate of patients. Non-small cell lung cancer (NSCLC) accounts for about 80% of all types of lung cancers, and lung adenocarcinoma (LUAD) accounts for about 40–50% of NSCLC. It has been reported that LUAD usually has some invasive subtypes. LUAD patients with epidermal growth factor receptor (EGFR) mutations have a response to anti-EGFR targeted therapy. Recently, Japanese scholars have studied the immunophenotype of LUAD patients with EGFR mutation in the tumor microenvironment (TME), where LUAD patients with EGFR mutation were treated by anti-programmed cell death protein 1 (PD-1) antibody monotherapy, while this method was basically ineffective. Therefore, targeted therapy was only effective for some appropriately selected patients with LUAD.

Considering the advantages of recognizing the immune response in solid tumors and the immunological changes during the immunotherapy of LUAD, immunotherapy has showed advantages in the treatment of lung cancer. Among them, PD-1 inhibitors that block the programmed cell death protein 1 (PD-1)/its ligand (PD-L1) pathway have long been the first-line treatment drugs for advanced non-small cell lung cancer with positive PD-L1 expression. This therapy significantly improved the prognosis of NSCLC patients and increased the 5-year survival rate of NSCLC patients to 15.5–23%. In a clinical trial, 305 patients with high expression of PD-1 were randomly enrolled, and were given pembrolizumab at a fixed dose of 200 mg every 3 weeks for 35 cycles or until disease progression, which were compared with chemotherapy drugs. The progression-free survival rate and overall response rate of chemotherapy drugs were significantly improved. Recent studies have shown that PD-1 is highly expressed in non-small cell lung cancer.

Therefore, it is of great significance to understand the relationship between the immune system and prognosis, which can help us make more effective use of immunoncology, promote the development of this discipline, and improve the efficacy of treating tumors. The Li’s research team was the first in China to propose a model based on immune score to predict the survival and the effects of adjuvant chemotherapy in patients with gastric cancer after surgery, and predict whether patients with stage II and stage III gastric cancer can benefit from adjuvant chemotherapy. In addition, Wang et al studied the application of immune score in the prognosis of breast cancer. However, studies on the prognosis of LUAD based on immune scores have not been reported yet. Therefore, this study aimed to evaluate the relationship between immune score and prognosis, and then based on this relationship, we developed a clinical nomogram of predicting the survival of LUAD patients. The details were reported as follows:

**Methods**

**Materials**

The data used in this study were downloaded from The Cancer Genome Atlas database (TCGA). TCGA is a project funded by the National Cancer Institute (NCI) and National Human Genome Research Institute (NHGRI) and has generated comprehensive and multi-dimensional maps of the key genomic changes in various cancers. TCGA is currently the largest public data set for tumor genome analysis, including more than 200 cancer and clinical information, as well as DNA methylation, gene expression, somatic mutation, etc. In April 2020, we downloaded the LUAD raw data from the TCGA database (http://www.cbioportal.org/).

The immune score data was collected from a public data platform (https://cancergenome.nih.gov/). The immune score was obtained based on the number of lymphocytes in the central area of the tumor and the infiltration junction area. The score suggested the degree of tumor immune cell infiltration.

**Data Preprocessing**

The downloaded LUAD case data and the corresponding immune score data were firstly preprocessed to remove cases with incomplete information, and finally the data of 411 LUAD cases were obtained. The R language software (version 3.6.3) was used to merge the two types of data, the duplicate cases were removed, and finally 407 cases were obtained, for whom there was a consistent one-to-one match between each case and each immune score. Only a very small number of patients were in the advanced stage, and there was no statistical significance. Therefore, the patients in the advanced stage were removed, and finally 390 cases were enrolled for analysis.

The data of cases in this study was processed as follows. The survival status was divided into two categories: 0 meant survival, and 1 meant death. Age was classified into 3 categories: 0 represented “≤50 years old”, 1 represented “>50 and
≤70 years old", and 2 represented “>70 years old”. The tumor TNM comprehensive stage was divided into 3 categories: 0 represented “I”, 1 represented “II”, and 2 represented “III”. Immune scores were classified into 3 categories: 0 for “low score (≤698.1)”, 1 for “medium score (>698.1 and ≤1246.3)”, and 2 for “high score (>1246.3)”. There were two categories of gender: 0 for “male”, and 1 for “female”. Whether or not to undergo radiotherapy was divided into two categories: 0 for “radiotherapy has been carried out”, 1 for “radiotherapy has not been carried out”. The details of each stage of data pre-processing and the sample size of each stage are shown in Figure 1.

In addition, the data sets were randomly divided into two groups, one was the training set, and the other was the validation set, which was used to do the decision curve analysis (DCA). The DCA is a simple method for evaluating clinical predictive models, diagnostic tests and molecular markers. The abscissa is the threshold probability, which represents the probability that the model predicts the occurrence of an event. The ordinate is the net benefit, which represents When various evaluation methods reach a certain value, the risk probability of patient i is denoted as Pi. When Pi reaches a certain threshold (Pt), it is defined as positive and some kind of intervention is adopted, which brings the benefits to the patients undergoing treatment, and the harms to the non-patients undergoing treatment and the patients without treatment. The ordinate is the net benefit of the benefits minus the harms.

**Statistical Analysis**

The downloaded data provided the overall survival (OS) time of each patient. OS time was defined as the time from pathological diagnosis to death. X-tile software (version 3.6.1, Yale University School of Medicine, New Haven, CT, USA) was used to get the cut-off point of the immune score, and X-tile was used to evaluate the immune score, and the immune score was divided into three groups, as shown in Figure 2.

![Figure 1 Research flow chart of the sample collection.](https://doi.org/10.2147/IJGM.S337934)
Results

 Patients’ Characteristics

The average age of the enrolled patients was 65.17 years old (SD = 10.09, Range 39–88 years). Twenty-nine patients (7.4%) were younger than 50 years old, 227 patients (58.2%) ranged from 51 to 70 years old, and 134 patients (34.4%) were older than 70 years old. Among the 390 patients, there were 172 (44.1%) males and 218 (55.9%) females. According to TNM classification, there were 223 (57.2%) patients in stage I, 105 (26.9%) patients were in stage II, and 62 (15.9%) patients were in stage III. Patients were divided into groups with high, medium, and low immune scores (Figure 2). A total of 143 (36.7%) patients with an immune score less than or equal to 698.1 were included in the low score group, 152 (39.0%) patients with an immune score greater than 1246.3 were included in the high score group, and 95 (24.4%) patients with an immune score greater than 698.1 and less than or equal to 1246.3 were included in the medium score group. Among them, 53 patients (13.6%) received radiotherapy, and 337 patients (86.4%) did not receive radiotherapy.

The clinicopathological characteristics of the study subjects in different immune score groups were listed in Table 1. The average age of the high-, medium- and low-score groups were 62.27 years old (SD = 10.48), 68.05 years old (SD = 9.39), and 66.24 years old (SD = 9.42), respectively. In the “age ≤50 years old” group, the number of patients with the low immune score was the largest, and so does the “age >51 and ≤70 years old” group. Patients who received radiotherapy were evenly distributed in the different immune scores groups. However, among the patients who did not receive radiotherapy, the proportion of patients with the high immune score was the highest. For the female patients, the proportion of patients with the high immune score was the highest, while for the male patients, the proportion of patients with the high immune score was the lowest. In the TNM staging of tumors, stage I and II were more likely to appear in patients with the high immune score.

Univariate and Multivariate Analysis of OS

The univariate analysis was performed to explore the correlation between the OS and clinicopathological characteristics in LUAD patients. As shown in Table 2 and Figure 3, patients with the low, medium, and high immune scores had significant differences in OS (P < 0.001). In addition, there was a statistically significant difference in...
OS of the different TNM stages of tumor (P <0.001). There was also a difference in OS of patients whether receiving radiotherapy or not (p = 0.0015). However, there were no differences in age and gender (p > 0.05).

The results of multivariate Cox proportional hazards regression analysis are shown in Table 3. Compared with patients with low and medium immune scores, the OS of patients with the high immune score was significantly
improved (HR and 95% CI: 0.489 [0.324–0.737]). For the tumor staging, the OS of patients in the stage II and III was significantly worse than those in the stage I, and HR and 95% CI for stages II and III were 2.167 [1.450–3.238] and 2.588 [1.636–4.094], respectively. Patients who did not receive radiotherapy had worse OS than those who received radiotherapy, with HR and 95% CI (0.640 [0.414–0.989]). Therefore, whether to receive radiotherapy or not, the TNM stage and the immune score were included in the multivariate model.

### Prognostic Nomogram of OS

Cox regression model was used to evaluate the associations between potential prognostic factors and OS by univariate and multivariate analysis. The potential prognostic factors included the immune score, tumor staging and whether to receive radiotherapy. According to the results of the multivariate Cox regression analysis, the forest curve was drawn. The forest plot showed the relationship between the potential prognostic factors and OS (Figure 4).

The prognostic nomogram of OS, which integrated all significant independent factors obtained from the multivariate analysis, is shown in Figure 5. The C-index for predicting OS was 0.691 (95% CI, 0.646–0.736). Based on the above analysis, a receiver-operator-characteristics (ROC) curve was drawn, and the area under the curve (AUC) was calculated. The AUC values of 3- and 5-year OS were 0.698 and 0.684, respectively (Figure 6). The reliability of the prediction model was verified by the C index and ROC curve.

The calibration curves of 3- and 5-year OS showed a good consistency in the probability between the actual observation and the nomogram prediction (Figure 7A and B).

### Decision Curve Analysis

In the DCA chart, there are four curves. The black None line represented that all patients did not have LUAD and did not receive any treatment, and the net benefit (Net Benefit) was 0. The gray All line represented that all patients had LUAD and received the treatment, and the net benefit was a backslash with a negative slope.

### Table 3 Multivariate Analyses of OS Among LUAD Patients According to Immune Scores and Clinic Pathological Characteristics

| Characteristics | Total | Survival | Death | HR (95% CI) | P value |
|-----------------|-------|----------|-------|-------------|---------|
| Radiation       |       |          |       |             |         |
| Yes             | 53    | 23 (43.4)| 30 (56.6)| 1.0         | 0.044   |
| No              | 337   | 231 (68.5)| 106 (31.5)| 0.640 (0.414, 0.989) |      |
| TNM stage       |       |          |       |             |         |
| I               | 223   | 167 (74.9)| 56 (38.6)| 1           | <0.001  |
| II              | 105   | 60 (57.1)| 45 (42.9)| 2.167 (1.450, 3.238) | <0.001  |
| III             | 62    | 27 (43.5)| 35 (56.5)| 2.588 (1.636, 4.094) | <0.001  |
| Immune scores   |       |          |       |             |         |
| ≤683.7          | 143   | 82 (57.3)| 61 (42.7)| 1           |         |
| 683.7 to 1246.3 | 95    | 57 (60.0)| 38 (40.0)| 0.828 (0.548, 1.251) | 0.371   |
| >1246.3         | 152   | 115 (75.7)| 37 (24.3)| 0.489 (0.324, 0.737) | <0.001  |
Figure 4 Forest plot of overall survival (OS). *P<0.05, ***P<0.001.

Figure 5 Nomograms for predicting survival of LUAD.
In the group of the training set, the blue Immune_score curve was above the black None line and the gray All line, when the high-risk threshold value ranged from 0.20 to 0.52, suggesting the model had clinical practical value, and the red Nomogram curve was above the black None line and the gray All line, when the high-risk threshold value ranged from 0.20 to 0.71, suggesting the model had clinical practical value (Figure 8A).

In the group of the validation set, the blue Immune_score curve was above the black None line and the gray All line, when the high-risk threshold value ranged from 0.24 to 0.36, suggesting the model had clinical practical value, and the red Nomogram curve was above the black None line and the gray All line, when the high-risk threshold value ranged from 0.12 to 0.60, suggesting the model had clinical practical value (Figure 8B).

The DCA of the two data sets both confirmed that the model constructed in this study had clinical practical value.

Discussion

In this study, publicly available data from the TCGA repositories was used to identify the immune score associated with OS of LUAD patients in the stages I, II, and III. After adjusting for possible confounding factors, we found that the high immune score was significantly associated with OS of LUAD patients. At the same time, a nomogram of predicting the survival time of LUAD patients was established.

The important contribution of immune cells to LUAD has been widely accepted, and immune genes were considered as biomarkers of immune response in immunotherapy. Early studies also showed that some immune genes were significantly related to the prognosis.
In addition, some studies suggested that immune gene expression should be included in the current multi-gene test to improve the prognosis of patients with LUAD. However, these findings have not been used in clinical studies to predict the probability of OS of LUAD. In addition, few studies established a nomogram using the immune score. In this study, the clinicopathological information and the immune scores of LUAD patients from the TCGA were used to explore their association with prognosis. Furthermore, a nomogram was established to make the prognostic assessment of LUAD patients easier.

In this study, after adjusting for several possible confounding factors, we found the higher the immune score was, the higher the OS of LUAD patients was. Similar results were observed in the study of Pagès et al. The possible reason might be that the high immune score meant that the immune system and immune function were enhanced, which could improve the anti-tumor immunity of the tumor microenvironment, thereby controlling and eliminating the tumor. In addition, related studies found that genes related to immune cell activation significantly increased in relatively long-lived patients with cancer. Moreover, some important genes, such as CD302, etc., were used to calculate immune characteristics and play important roles in immune function. Furthermore, a study on T cell-related markers indicated that LUAD patients with higher expression of CD3+ and CD8+ receiving relevant neoadjuvant chemotherapy had a good prognosis. Therefore, the immune score could not only be used as a biomarker for the prognosis of LUAD patients but also has a potential clinical value in the selection of treatment strategies.

In this study, we found that whether to receive radiotherapy was an important independent prognostic factor for patients with LUAD (p = 0.0015). Patients who received radiotherapy had better OS, with HR: 0.517 and 95% CI: 0.344–0.776. With the improvement of the awareness of early cancer screening among citizens and the level of medical and health care, the early detection rate of cancer increased significantly, which was similar to our study in which more than 50% of the patients enrolled were in the stage I. However, our results were different from previous studies. The previous studies suggested that most of the LUAD patients were in the advanced stage (III/IV) at the time of initial diagnosis, and the best opportunity for surgery had been missed. Early detection greatly improves the success rate of the patients’ treatment and prolongs the patients’ survival time. For LUAD patients in the early stage, surgery is the main treatment supplemented by adjuvant radiotherapy and chemotherapy. However, some studies reported that long-term chemical/radiotherapy had the possibility of carcinogenesis. Postoperative radiotherapy and chemotherapy can reduce the patient’s own immunity and weaken the defense against the tumor, which is conducive to the tumor regeneration. For LUAD patients with high immune scores before and after surgery, immunotherapy or combined immunotherapy based on precise gene sequencing can reduce the harm of radiotherapy and chemotherapy. Studies reported that postoperative immunotherapy could generally reduce the local recurrence rate by about 30%, significantly reduce distant metastases, and increase the 5-year survival rate by
about 20%. Even if the tumor recurred, the recurrence time could be obviously delayed. It was confirmed that for treating the resectable non-small cell lung cancer, neoadjuvant immunotherapy was highly safe and did not affect surgery, with 45% of the pathologically significant remission rate and 73% of the 18-month recurrence-free survival rate. Studies have also shown that for the early-stage non-small cell cancer, immune drugs can block a variety of molecular markers to kill cancer cells with fewer side effects. Therefore, in addition to the traditional lung cancer treatment, immunotherapy can improve the survival rate of patients.

The LUAD prognostic model was constructed based on the immune cell infiltration score and clinicopathological characteristics, clarifying the relationship between immune cell infiltration and the occurrence and development of LUAD. The prognostic model we established could effectively predict the 3- and 5-year survival rates of LUAD patients, which suggested the roles of different immune scores in the development of LUAD. These findings might provide new ideas for the treatment and prognosis of LUAD from the perspective of immune cell infiltration.

However, there are still some limitations in our study. There were relatively few gene expression data sets used to calculate immune scores. The LUAD patients in the stage IV were not enrolled in this study. We only made the statistical analysis of the treatment of whether to receive the radiotherapy or not and did not analyze the other treatment methods for LUAD patients. No external data verification was performed. In addition, the nomogram is a little simple and has a limited effect in the accurate prediction of the survival of patients, and thus in the future, the more comprehensive prognostic model will be established to improve the accuracy of predicting the survival of patients. For us, there is still a lot of work to do in the future. In addition to clinical pathological factors, further efforts will be made to collect case data related to immune gene expression, thereby updating and developing our model.

Conclusions
This study found that for LUAD patients in the stages I, II, and III, the high immune score was significantly associated with better OS. In addition, the established nomogram could effectively predict the prognosis of LUAD. This practical prognostic model can easily evaluate the OS of patients and identify the patients who need active treatment.

Abbreviations
NSCLC, non-small cell lung cancer; OS, overall survival; LUAD, lung adenocarcinoma; HR, hazard ratios; CI, confidence intervals; c-index, concordance index; EGFR, epidermal growth factor receptor; TME, the tumor microenvironment; PD-1, programmed cell death protein 1; PD-L1, its ligand (PD-L1); TCGA, The Cancer Genome Atlas; NCI, National Cancer Institute; NHGRI, National Human Genome Research Institute; ROC, receiver-operator-characteristics; AUC, area under the curve.

Data Sharing Statement
The datasets used and/or analysed during the current study are available from the corresponding author on reasonable request.

Ethics Approval and Consent to Participate
Not applicable, because TCGA belongs to public databases, the patients involved in the database have obtained ethical approval, users can download relevant data for free for research and publish relevant articles, and our study is based on open-source data, and the Affiliated Hospital (Clinical College) of Xiangnan University do not require research using publicly available data to be submitted for review to their ethics committee, so there are no ethical issues and other conflicts of interest.

Acknowledgments
Thanks for the support from the funders and TCGA database.

Funding
This study was supported by: 1. Science and Technology Funding Project of Hunan Province China (No.2017SK4010). 2. Key Laboratory of Tumor Precision Medicine, Hunan colleges and Universities Project (2019-379). 3. Science and Technology Project of Chenzhou city (No. CZJK2016031). The funding bodies played no role in the design of the study and collection, analysis, and interpretation of data and in writing the manuscript, but paid our wages, and thus we can concentrate on the studies to get more achievements.

Disclosure
The authors declare that they have no conflicts to disclose.
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