Construction and Verification of a Predictive Model for Risk Factors in Children With Severe Adenoviral Pneumonia

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Objective: To construct and validate a predictive model for risk factors in children with severe adenoviral pneumonia based on chest low-dose CT imaging and clinical features.

Methods: A total of 177 patients with adenoviral pneumonia who underwent low-dose CT examination were collected between January 2019 and August 2019. The assessment criteria for severe pneumonia were divided into mild group (N = 125) and severe group (N = 52). All cases divided into training cohort (N = 125) and validation cohort (N = 52). We constructed a prediction model by drawing a nomogram and verified the predictive efficacy of the model through the ROC curve, calibration curve and decision curve analysis.

Results: The difference was statistically significant (P < 0.05) between the mild adenovirus pneumonia group and the severe adenovirus pneumonia group in gender, age, weight, body temperature, L/N ratio, LDH, ALT, AST, CK-MB, ADV DNA, bronchial inflation sign, emphysema, ground glass sign, bronchial wall thickening, bronchiectasis, pleural effusion, consolidation score, and lobular inflammation score. Multivariate logistic regression analysis showed that gender, LDH value, emphysema, consolidation score, and lobular inflammation score were severe independent risk factors for adenovirus pneumonia in children. Logistic regression was employed to construct clinical model, imaging semantic feature model, and combined model. The AUC values of the training sets of the three models were 0.85 (0.77–0.94), 0.83 (0.75–0.91), and 0.91 (0.85–0.97). The AUC of the validation set was 0.77 (0.64–0.91), 0.83 (0.71–0.94), and 0.85 (0.73–0.96), respectively. The calibration curve fit good of the three models. The clinical decision curve analysis demonstrates the clinical application value of the nomogram prediction model.
INTRODUCTION

Adenovirus (ADV) pneumonia is one of the more common community-acquired severe pneumonia in children. Adenovirus infections are more common in young children, specifically in 6-month to 2-year-old infants owing to a lack of humoral immunity (1). Severe adenoviral pneumonia has a rapid onset, severe disease, rapid progress, and a poor prognosis (2). Severe adenoviral pneumonia has a high child mortality rate and a long recovery period, easily linked to multiple organ damage. The mortality rate of untreated severe adenovirus pneumonia or other communicable diseases may exceeds 50%. Severe lung lesions, acute respiratory distress syndrome (ARDS), and severe coagulation abnormalities are the direct causes of death in children. 14–16% of recovered patients may have different degrees of sequelae (3). Occlusion Bronchitis and bronchiectasis are the most common complications in children with severe adenoviral pneumonia. We have encountered some patients with adenoviral pneumonia with mild clinical symptoms who rapidly develop respiratory failure in our clinical practice. Consequently, identifying risk factors early and reminding clinicians to implement intervention strategies as soon as possible is critical. The imaging changes of children with adenovirus pneumonia appear early. The imaging examinations are crucial for diagnosing and condition judgment of adenovirus pneumonia. At present, clinical radiology diagnosis reports primarily rely on assessing the severity of lung infiltration, which is highly subjective and lacks quantitative parameter indicators and related standards. This study aimed to use the lung consolidation scores, lobular inflammation scores, and assessment of small airway lesions, combined with clinical data and laboratory examinations, to establish a predictive model for risk factors with severe adenovirus pneumonia in children.

MATERIALS AND METHODS

Patient Information

The study protocol was approved by the Ethical Review Committee of Hunan People's Hospital. The parents or guardians of all of the participants in this cooperative study provided written informed consent. This study retrospectively analyzed 721 patients with adenovirus pneumonia hospitalized in Hunan Provincial People's Hospital from January 2019 to December 2019. They were excluded if they had congenital lung dysplasia, airway malformations, congenital immunodeficiency, congenital heart disease, malnutrition, or congenital metabolic syndrome disease and 177 children were enrolled after screening (Figure 1).

Conclusion: The prediction model based on chest low-dose CT image characteristics and clinical characteristics has relatively clear predictive value in distinguishing mild adenovirus pneumonia from severe adenovirus pneumonia in children and might provide a new method for early clinical prediction of the outcome of adenovirus pneumonia in children.

Keywords: children, adenovirus pneumonia, low-dose CT, nomogram, prediction model
FIGURE 1 | Workflow of the study. Workflow can be divided into three parts: patient selection, feature extraction, and model construction.

or Philips 256 Brilliance iCT (Philips, Netherlands). The voltage was 100 KV, the tube current was 30–50 mA, the reconstruction layer thickness was 1.25 mm, and the layer spacing was 1.25 mm.

CT Image Analysis
Two senior diagnosticians (with 21 and 11 years of experience in chest imaging diagnosis) evaluated all imaging data. The standard of chest CT examination results was based on the priority definition of the “Fleischner Society Glossary” (6), including assessment of lung consolidation and extent of involvement, lobular inflammation and extent of involvement, assessment of small airway disease, including emphysema, mosaic sign, bronchial wall thickening, and bronchiectasis, and whether there is pleural effusion, pneumothorax, emphysema mediastinum, and other extrapulmonary complications. On the CT image, the lungs were divided into six regions: upper (above the carina), middle (below the carina to above the inferior pulmonary vein), and inferior (below the inferior pulmonary vein) (7, 8). Each area was classified as lung consolidation or lobule inflammation involvement range assignment (9). Each area was scored from 0 to 4 to evaluate the degree of pneumonia involvement in each lung area; 0 was normal, four was the area where all the lung parenchyma was involved. The values of the six areas were accumulated, and the total CT score was between 0–24.

Statistical Methods
This study’s measurement data have been expressed as mean ± standard deviation or median (upper quartile, lower quartile). A chi-square test or Fisher’s exact test has been used for categorical variables and continuous variables that conform to a normal distribution and have been used in t-test, Mann–Whitney test for abnormally distributed continuous variables, univariate analysis of the influencing factors of imaging parameters, multivariate logistic analysis of clinical signs and laboratory test indicators, and multivariate analysis of independent influencing factors, according to the stepwise selection method. The selected method has been incorporated into multivariate analysis. The receiver operating characteristic (ROC) curve has been used to determine the performance of the machine learning model, and the accuracy, sensitivity, specificity, and area under the curve (AUC) are calculated. All statistical analyses were performed using IPMs (GE Healthcare, United States) R (version 3.5.1). A two-tailed p-value < 0.05 indicated statistical significance.

The imaging and clinical data of the mild and severe groups were analyzed and compared, using univariate analysis to extract the selected imaging and clinical features. Multivariate logistic analysis was used to conduct multivariate analysis of independent influencing factors. The selected method was incorporated into a multi-factor analysis to screen out the image characteristics and clinical characteristics. Finally, these characteristics were used to establish a prediction model for severe childhood adenovirus pneumonia patients using logistic regression. The risk factors were combined to construct a predictive model, and a nomogram was developed using R language rms package. The enrolled patients were divided into a prediction group and a verification group, and the receiver operating characteristic (ROC) curve was used to determine the performance of the machine learning model, calculate the accuracy, sensitivity, specificity, and area under the curve (AUC), and test the stability of the prediction model for sex and reliability. The calibration curve was used to detect the fit of the model, and DCA evaluated the clinical benefit of patients (Figure 1).

RESULTS

General Situation
Table 1 shows the detailed information of patients. Among the 177 patients enrolled in the group, 125 (70.6%) were in the mild group and 52 (29.4%) were in the severe group; 149 were males, including 117 (78.5%) in the mild group and 32 (21.5%) in the severe group; 28 were females, including 8 (28.6%) in the mild group and 20 (71.4%) in the severe group. The number of male cases was significantly higher than that of females, but most of them were in the mild group, and fewer progressed to the severe group; there were fewer female cases in the mild group, but more progressed to the severe group. The average age of onset in the
mild group was 36 months (18.1, 55.3), 12 months longer than the average age of onset in the severe group (12, 36). The average weight of the mild group was 15 kg (11.0, 18.1), which was higher than the average weight of the severe group by 12 kg (9.2, 14.4).

Clinical Signs and Laboratory Examinations
All the patients in the group suffered from fever. The average body temperature of the mild group was 37.1°C (36.6, 38.3), which was lower than the severe group’s 37.9°C (37.0, 38.7); the fever duration between the mild and severe groups was 5 days, and there was no significant difference (Table 1).

In the mild disease group, the lymphocyte/neutrophil ratio (L/N ratio), lactate dehydrogenase (LDH), alanine transaminase (ALT), asparaginase aspartate aminotransferase (AST), creatine kinase-MB form (CK-MB), and ADV DNA levels, as well as White Blood Cell Count (WBC) and neutral hemoglobin, were significantly lower than those in the severe group. There was no significant difference in the levels of neutrophil (NEU), lymphocyte, lymphocyte/white Blood Cell ratio (L/W ratio) and C-reactive protein (CRP) (Table 1).

Imaging Performance
Through the analysis of differences, meaningful features have been identified. The symptoms of bronchial inflammation, emphysema, ground glass sign, bronchial wall thickening, bronchiectasis, and pleural effusion in the mild group differed from those in the severe group (Table 1). The consolidation score of 0 (0, 1) in the mild group was significantly lower than the consolidation score of 3 (1, 4) in the severe group ($P < 0.001$). The lobular inflammation score of the mild group was significantly lower than the severe group ($P = 0.014$). A total of 78 patients in the enrolled group had bronchial inflammation signs (Figure 2A); among them, 46 patients (36.8%) were in the mild group and 32 patients (61.5%) were in the severe group. 85 patients had emphysema (Figure 2B). Among them, 48 cases were in the mild group (38.4%) and 37 cases were in the severe group (71.2%). The incidence of emphysema in the mild group was significantly lower than the severe group ($P < 0.001$). 44 patients had a mosaic sign (Figure 2C). Among them, 33 patients (26.4%) were in the mild group and 11 patients (21.2%) were in the severe group. 150 patients had bronchial wall thickening (Figure 2D); among them, 101 patients (80.8%) were in the mild group and 49 patients (94.2%) were in the severe group. 8 patients had bronchiectasis (Figure 2E), of which five patients (4.0%) were in the mild group and 3 patients (5.8%) were in the severe group. 12 patients had pleural effusion (Figure 2F); among them, the mild group had 4 cases (3.2%) and 8 cases (15.4%) were in the severe group. The incidence of pleural effusion in the mild group was significantly lower than that in the severe group ($P = 0.009$).

Multivariate Logistic Regression Analysis
A multivariate logistic analysis analyzed clinical signs, laboratory test indicators, and independent factors. The multivariate analysis was incorporated according to a stepwise selection method. The results showed that gender, LDH value, emphysema, consolidation score, and lobular inflammation score were independent risk factors for predicting adenovirus pneumonia in children with a severe health condition ($all P < 0.05$) (Table 2).

Establishment and Verification of Nomogram Model
The nomogram (Figure 3) shows that the risk score of gender (1 male, 0 female) was 18 for females, and the risk score for pleural effusion was 24. The risk score of LDH value, consolidation score, and lobular inflammation score varied with children’s glands. The actual value of patients with viral pneumonia increased.

We have established three different models: clinical model, imaging model, and clinical + imaging combined model (Table 3). The results show that the AUC of the clinical model, imaging model, and combined model in the training set was 0.85, 0.83, and 0.91, respectively (Figure 4A); The AUCs of the clinical model, imaging model, and combined model of the validation set were 0.77, 0.83, and 0.85, respectively (Figure 4B).

According to the calibration graphs of the training queue (Figure 5A) and the verification queue (Figure 5B), the standard curve in the calibration graph fits well with the predicted calibration curve, and the degree of calibration is good. Hosmer test $P$-value is greater than 0.05, and there is no statistical difference.

The analysis of clinical decision curve (Figure 6) shows that the nomogram prediction model has good predictive ability, making appropriate clinical decisions for early severe childhood adenovirus patients. The combined model can bring the greatest clinical benefit to patients.

DISCUSSION
Human adenovirus (HAdV) was first discovered in 1953 by Rowe and colleagues. It is a non-enveloped double-stranded DNA virus belonging to the genus human mammary adenovirus. They are common pathogens in children and are responsible for various diseases (10). Among the 90 AdV serotypes, 55 are known to cause human diseases, of which AdV3, 4, 7, and 14 are the most common types that cause respiratory disease outbreaks (11). However, in most cases, HAdV is associated with mild to moderate disease, some patients may still develop life-threatening diseases, especially those with a weakened immune system (12). Adenovirus is primarily transmitted through air droplets. Infants, children, the elderly, and people with weakened immune systems are more likely to be infected and develop severe pneumonia (13). Infants and young children are the main populations, and the incidence usually occurs under the age of 2. The severe group in this study had a younger onset age, with an average age of 12 months, which was significantly lower than the mild group ($P < 0.001$). Children in the severe group weighed significantly less than those in the mild group ($P < 0.001$). There are more boys than girls in this study population, which is consistent with the results of other scholars (14, 15). In this study, the proportion of girls who progressed to severe illness was significantly higher than boys, which may be related to the girls’ low weight, nutritional status, and weak
# TABLE 1 | Demographic, clinical and chest CT findings of patients with adenovirus pneumonia included in the study.

| Characteristics                  | Total  | MAP group  | SAP group  | P-valuea | Training Cohort | Validation Cohort | P-valueb |
|----------------------------------|--------|------------|------------|----------|-----------------|-------------------|----------|
|                                  | (n = 177) | (n = 125) | (n = 52) |          | (n = 125) | (n = 52) |          |
| Demographics                     |        |            |            |          |                 |                   |          |
| Gender                           |        |            |            | <0.001   | 0.919           |                   |          |
| Male                             | 149    | 117        | 32         |          |                 |                   |          |
| Female                           | 28     | 8          | 20         | <0.001   | 105 (12.48)     | 24 (12.48)        | 0.908    |
| Age (month)                      | 177    | 36 (18.1, 55.3) | 12 (12.36) | <0.001   | 36 (15.0, 16.7) | 13 (10.7, 17.9)  | 0.914    |
| Weight                           | 177    | 15 (11.0, 18.1) | 12 (9.2, 14.4) | <0.001   | 14 (10.5, 16.7) | 13 (10.7, 17.9)  | 0.914    |
| Clinical features                |        |            |            |          |                 |                   |          |
| Fever (°C)c                      | 177    | 37.1 (36.6, 38.3) | 37.9 (37.0, 38.7) | <0.001   | 37.4 (36.8, 38.4) | 37.2 (36.6, 38.5) | 0.715    |
| Fever course (days)d             | 177    | 5 (4, 7) | 5 (4, 7) | 0.988    | 5 (4, 7) | 5 (4, 6) | 0.895    |
| Laboratory detection             |        |            |            |          |                 |                   |          |
| WBC (×10^9/L)                    | 177    | 7.58 (5.17, 10.31) | 6.32 (4.8, 9.6) | 0.169    | 6.67 (4.94, 9.99) | 7.54 (5.38, 10.06) | 0.629    |
| NEU                              | 177    | 3.34 (2.33, 5.38) | 4.57 (3.17, 8.01) | 0.005    | 3.55 (2.63, 6.57) | 3.62 (2.37, 5.91) | 0.842    |
| Lymphocyte                       | 177    | 2.73 (1.83, 3.84) | 1.98 (1.33, 4.54) | 0.168    | 2.68 (1.69, 4.14) | 2.34 (1.66, 3.53) | 0.66     |
| L/N ratio                        | 177    | 0.66 (0.48, 1.26) | 0.48 (0.35, 8.4) | 0.002    | 0.65 (0.36, 1.05) | 0.66 (0.47, 1.17) | 0.591    |
| L/W ratio                        | 177    | 0.36 (0.29, 0.50) | 0.32 (0.25, 0.51) | 0.36     | 0.36 (0.27, 0.51) | 0.36 (0.29, 0.49) | 0.929    |
| CRP                              | 177    | 12.8 (4.3, 28.3) | 12.5 (4.1, 18.9) | 0.3      | 12.60 (4.33, 24.14) | 15.70 (4.06, 33.81) | 0.213    |
| LDH                              | 177    | 343 (285, 425) | 629 (423, 989) | <0.001   | 393 (306, 605) | 34 (298, 509) | 0.187    |
| ALT                              | 177    | 15.7 (11.8, 21.3) | 22.1 (16.5, 32.0) | <0.001   | 16.9 (12.8, 25.2) | 18.1 (11.7, 24.0) | 0.766    |
| AST                              | 177    | 40.9 (31.1, 58.1) | 66.2 (47.3, 102.6) | <0.001   | 46.9 (34.9, 69.8) | 43.6 (32.5, 63.9) | 0.386    |
| CK-MB                            | 177    | 24 (18, 32) | 38 (23, 53) | <0.001   | 28 (20, 41) | 23.5 (17.5, 34.7) | 0.100    |
| ADV DNA levele                   | 177    | 5.76 (3.96, 7.26) | 6.62 (5.66, 7.58) | 0.017    | 6.31 (4.65, 7.40) | 5.58 (3.92, 7.07) | 0.07     |
| Imaging characteristics          |        |            |            |          |                 |                   |          |
| Air bronchogram                  |        |            |            | 0.003    | 0.193           |                   |          |
| Positive                         | 78     | 46         | 32         | 59       | 19              |                   |          |
| Negative                         | 99     | 79         | 20         | 66       | 33              |                   |          |
| Pulmonary emphysema              |        | <0.001*    |            |          |                 |                   |          |
| Positive                         | 85     | 48         | 37         | 64       | 21              |                   |          |
| Negative                         | 92     | 77         | 15         | 61       | 31              |                   |          |
| Mosaic sign                      |        | 0.462      |            |          |                 |                   | 0.978    |
| Positive                         | 44     | 33         | 11         | 31       | 13              |                   |          |
| Negative                         | 133    | 92         | 41         | 94       | 39              |                   |          |
| Bronchial wall thicken           |        | 0.024      |            |          |                 |                   | 0.005    |
| Positive                         | 150    | 101        | 49         | 112      | 38              |                   |          |
| Negative                         | 27     | 24         | 3          | 13       | 14              |                   |          |
| Bronchiectasis                   |        | 0.608      |            |          |                 |                   | 0.905    |
| Positive                         | 8      | 5          | 3          | 5        | 3               |                   |          |
| Negative                         | 169    | 120        | 49         | 120      | 49              |                   |          |
| Pleural effusion                 |        | 0.009*     |            |          |                 |                   | 0.987    |
| Positive                         | 12     | 4          | 8          | 9        | 3               |                   |          |
| Negative                         | 165    | 121        | 44         | 116      | 49              |                   |          |
| Consolidation score              | 177    | 0 (0, 1)   | 3 (1, 4)   | <0.001*  | 1.00 (0.00, 2.00) | 0.00 (0.00, 2.00) | 0.13     |
| Lobular inflammation score       | 177    | 3 (1, 5)   | 4 (2, 6)   | 0.014*   | 3.00 (1.00, 5.00) | 2.00 (1.00, 4.00) | 0.051    |

Categorical variables are presented as number of patients and data in parentheses are percentages; continuous variables are showed as median (lower quartile, upper quartile).

Statistical analysis was performed with Wilcoxon Test or Fisher exact test. The symbol * for P < 0.05 suggests a significant difference.

- Comparing the group with ARDS and the group without ARDS.
- Comparing between the training cohort and the validation cohort.
- The temperature of fever was obtained at admission.
- Fever course was defined as the duration from onset to the day of admission.
- The data of ADV DNA level underwent a transition of log function.

MAP, mild adenovirus pneumonia; SAP, severe adenovirus pneumonia; WBC, white blood cell; NEU, neutrophile; CRP, C-reactive protein; LDH, lactate dehydrogenase; ALT, alanine transaminase; AST, aspartate transaminase; CK-MB, creatine kinase-MB.
immunity. We also cannot rule out the possibility of bias due to the small sample size.

All patients enrolled in this study suffered from fever. The average body temperature of the mild group was 37.1°C (36.6, 38.3), which was lower than the severe group’s 37.9°C (37.0, 38.7); coughing is the most common symptom in patients with adenovirus pneumonia in this group of cases, but it is a non-specific symptom. Fever is also a prominent symptom of this group of cases, and it is mostly a high fever. Adenovirus pneumonia is one of the few viral types of pneumonia that can cause an increase in white blood cell count. There was no significant difference in heat duration between the mild and severe groups, which may be related to the fact that we only counted the data within 3 days of the admission.

The laboratory test indexes L/N ratio, LDH, ALT, AST, CK-MB, and ADV DNA viral load of the mild group were significantly lower than those of the severe group, which may be related to the weak immunity of younger children; another possible mechanism is that adenovirus may have triggered a cytokine storm (9). The levels of AST and LDH in the severe group were significantly higher than those in the mild group, consistent with the findings of Huang et al. (15, 16). Children with severe adenovirus pneumonia may have pulmonary ventilation disorders, resulting in hypoxemia, severe microcirculation disorder, insufficient tissue perfusion, and abnormal cell metabolism. At the same time, pathogenic microorganisms and various inflammatory factors cause various levels of damage to hepatocytes and cardiomyocytes, increasing cell membrane permeability and the concentration of AST and LDH in serum (17), suggesting that adenovirus-induced inflammatory storms are crucial for disease transformation. It is also of great significance.

In this study’s univariate regression analysis, bronchiectasis, emphysema, bronchial wall thickening, pleural effusion, consolidation score, and lobular inflammation score in the chest low-dose CT findings high-risk factors for severe adenovirus pneumonia, while multivariate regression analysis found that,

![FIGURE 2 | Chest CT imaging findings of adenovirus pneumonia. (A) Right upper lung consolidation with bronchial inflation sign; (B) left emphysema; (C) double lung mosaic sign; (D) right lower lung bronchial wall thickening and surrounding lobule inflammation; (E) right Middle lung bronchiectasis; (F) a small amount of pleural effusion on the right.](image)

**TABLE 2 | Multivariable logistic regression for nomogram construction.**

| Characteristics                  | Odd ratio | 95% CI      | P-value |
|----------------------------------|-----------|-------------|---------|
| Gender                           | 0.21      | 0.05–0.83   | 0.026*  |
| LDH                              | 1.00      | 1.00–1.00   | 0.013*  |
| Pulmonary emphysema              | 8.28      | 2.23–30.75  | 0.001*  |
| Consolidation scores             | 1.53      | 1.08–2.17   | 0.017*  |
| Lobular inflammation scores      | 1.14      | 0.97–1.34   | 0.001*  |

*Male was denoted as 1, and female as 0.
The odd ratio was 0.21 means that male showed lower likelihood of severe pneumonia.

P Pulmonary emphysema positive was denoted as 1, and negative as 0.
The Odds Ratio was 8.28 means that patients with pulmonary emphysema showed higher likelihood of severe pneumonia.

*P value < 0.05, which showed significance.

![FIGURE 3 | Nomogram. To draw an upward vertical line to the “Points” bar to calculate points. Based on the sum, draw a downward vertical line from the “Total Points” line to calculate the probability of classification of adenoviral pneumonia patient. Male was denoted as 1, and female as 0.](image)
Emphysema, consolidation score, and lobular inflammation score were independent risk factors. Shen et al. (18) found that children with adenovirus pneumonia with pleural effusion had more severe clinical symptoms, consistent with our research results. Adenovirus and inflammatory mediators cause bronchi and bronchiole mucosal edema, hyperemia, neutrophils, lymphocytes, and other inflammatory cell infiltration around the bronchi and bronchiole and the wall of the tube, resulting in thickening of the bronchial tube wall and necrosis of the bronchial and bronchial mucosa. Necrosis obstructs the lumen as it falls off, and at the same time, mucus secretion increases and obstructs the lumen, resulting in emphysema. The centripetal distribution of lung consolidation after adenovirus pneumonia is closely related to its pathological changes. The lung parenchyma is affected by adenovirus infection, spreading through the small and medium bronchus along the airway. When the lesion
involves the bronchioles, it quickly spreads to the alveoli, causing the formation of inflammatory substances. Pulmonary consolidation suggests that many lesions and severely impaired lung function are the fundamental factors leading to severe adenovirus pneumonia.

Although immunofluorescence tests or nucleic acid tests are the standards for clinical diagnosis of adenovirus infection, its application still has some limitations. Not all medical institutions can carry it out. Chest CT imaging is an effective auxiliary diagnosis method that has been widely conducted. Experienced radiologists generate semantic features of different types of information about pneumonia from CT images of patients, such as lobular inflammation, consolidation, mosaic sign, bronchial wall thickening, bronchiectasis, emphysema, pleural effusion, etc. These features can provide useful information for the prognosis and diagnosis of pneumonia. These unique features can be qualitative descriptors, called semantic features that describe the shape and internal structure of the lesion and are scored by radiologists (19–21). They are classified depending on the radiologist’s visual assessment, limiting the lesion’s description to the range visible with the naked eye (22–24).

Semantic features have been used for the first time to construct monographs in this study, especially the consolidation scores and lobular inflammation scores have been used to construct a predictive model for severe childhood adenovirus pneumonia. Compared with imaging omics features, quantitative semantic features analysis is simpler and easier. First, if a large number of inflammatory infiltrations occur on both sides of the lung, it is challenging to outline the area of interest on the CT image accurately, and some patients with mild adenovirus pneumonia have no apparent abnormalities on the CT scans. Secondly, semantic features do not require deep learning software and analysis packages for complex data analysis and processing and construct monographs. The results are stable and reliable, and the clinical diagnosis value is great. The combined model has the highest diagnostic efficiency and can assist clinical diagnosis and prediction. It can be promoted in the majority of primary hospitals.

To construct this prediction model, we list strict including and excluding criteria. All the patients including were required to be diagnosed as mild adenoviral pneumonia at the first 3 days of admission. Therefore, 92 patients were excluded for the diagnosis of severe adenoviral pneumonia at that time. Meanwhile, as one of the best pediatric centers in our province, most patients with mild adenovirus pneumonia in our hospital have a favored prognosis, and only a rather small part of them might convert to severe adenoviral pneumonia. This prediction model shows that the image model has an AUC of 0.83 in the training set and 0.83 in the test set; the AUC of the clinical model in the training set is 0.85, and the AUC in the test set is 0.77; the image + clinical combined model has an AUC of 0.91 in the training set. The centralized AUC is 0.85. The image feature has a large weight and contribution value in this prediction model, with objective indicators and strong stability. The calibration curve in the calibration graph closely fits the prediction calibration curve good. The analysis of the clinical decision curve shows that the nomogram prediction model has an excellent predictive ability, which helps to make appropriate clinical decisions for early severe childhood adenovirus patients and has crucial clinical guidance significance.

This study has several limitations. First, the samples size of this study is limited. It is a single-center retrospective study. Only patients with adenovirus pneumonia undergoing CT examination were selected. The results may have certain deviations. For future studies, it will be necessary to increase the sample size and conduct long-term multi-center clinical research.

CONCLUSION

The prediction model based on chest low-dose CT image characteristics and clinical characteristics has relatively clear predictive value in distinguishing mild adenovirus pneumonia from severe adenovirus pneumonia in children. All the selected features of this prediction model are objective and easily-obtained variables, and our results might provide a new method for early clinical prediction of the outcome of adenovirus pneumonia in children.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

ETHICS STATEMENT

Written informed consent was obtained from the individual(s), and minor(s)’ legal guardian/next of kin, for the publication of any potentially identifiable images or data included in this article.

AUTHOR CONTRIBUTIONS

PL designed the research study. LX and SZ performed the research. HL contributed to the new reagents and analytic
tools. PL and YH analyzed the data and wrote the manuscript. All authors have read and approved the final manuscript.

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**REFERENCES**

1. Kajon A, Lynch J. Adenovirus: epidemiology, global spread of novel serotypes, and advances in treatment and prevention. *Semin Respir Crit Care Med.* (2016) 375:586–602. doi: 10.1055/s-0036-1584923
2. Xie L, Zhang B, Zhou J, Huang H, Zeng S, Liu Q, et al. Human adenovirus load in respiratory tract secretions are predictors for disease severity in children with human adenovirus pneumonia. *Virol J.* (2018) 15:123. doi: 10.1186/s12985-018-1037-0
3. Yu Z, Zeng Z, Zhang J, Pan Y, Chen M, Guo Y, et al. Fatal community-acquired pneumonia in children caused by re-emergent human adeno-virus 7d associated with higher severity of illness and fatality rate. *Sci Rep.* (2016) 6:37216. doi: 10.1038/srep37216
4. National Health Commission of the People's Republic of China, State Administration of Traditional Chinese Medicine. Guideline for diagnosis and treatment of community-acquired pneumonia in children (2019 version). *Chin J Clin Infect Dis.* (2019) 12:6–13.
5. National Health Commission of the People's Republic of China, State Administration of Traditional Chinese Medicine. Guideline for diagnosis and treatment of adenovirus pneumonia in children (2019 version). *Chin J Clin Infect Dis.* (2019) 12:161–6.
6. Hansell DM, Bankier AA, MacMahon H, McLoud TC, Müller NL, Remy J, Fleischner society: glossary of terms for thoracic imaging. *Radiology.* (2008) 246:697–722.
7. Das KM, Lee EY, Enani MA, Alljawder SE, Singh R, Bashir S, et al. CT correlation with outcomes in 15 patients with acute Middle East respiratory syndrome coronavirus. *AJR Am J Roentgenol.* (2015) 204:736–42. doi: 10.2214/AJR.14.13671
8. Ooi GC, Khong PL, Müller NL, Yiu WC, Zhou LJ, Ho JC, et al. Severe acute respiratory syndrome: temporal lung changes at thin-section CT in 30 patients. *Radiology.* (2004) 230:836–44. doi: 10.1148/radiol.2303030853
9. Cha MJ, Chung MJ, Lee KS, Kim TJ, Kim TS, Chong S, et al. Clinical features and radiological findings of adenovirus pneumonia associated with progression to acute respiratory distress syndrome: a single center study in 19 adult patients. *Korean J Radiol.* (2016) 17:940–9. doi: 10.3348/kjr.2016.17.6. 940
10. Radke JR, Cook JL. Human adenovirus infections: update and consideration of mechanisms of viral persistence. *Curr Opin Infect Dis.* (2018) 31:251–6. doi: 10.1097/QCO.0000000000000453
11. Prusinkiewicz MA, Mymryk JS. Metabolic reprogramming of the host cell by human adenovirus infection. *Virology.* (2019) 11:141–62. doi: 10.3390/v11020141
12. Lion T. Adenovirus infections in immunocompetent and immunocompromised patients. *Clin Microbiol Rev.* (2014) 27: 441–62.
13. Pförtmueller CA, Barbani MT, Scheffold JC, Hage E, Heim A, Zimmerli S. Severe acute respiratory distress syndrome (ARDS) induced by human adenovirus B21: report on 2 cases and literature review. *J Crit Care.* (2019) 51:99–104. doi: 10.1016/j.jcrc.2019.02.019
14. Li M, Han XH, Liu LY, Yao HS, Yi LL. Epidemiological characteristics, clinical characteristics, and prognostic factors of children with atopic hospitalized with adenovirus pneumonia. *BMC Infect Dis.* (2021) 21:1051. doi: 10.1186/s12879-021-06741-0

**SUPPLEMENTARY MATERIAL**

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fped.2022.874822/full#supplementary-material

15. Xu N, Chen P, Wang Y. Evaluation of risk factors for exacerbations in children with adenoviral pneumonia. *Biomed Res Int.* (2020) 2020:4878635. doi: 10.1155/2020/4878635
16. Huang H, Chen Y, Ma LY, Yan MM, Deng Y, Zhang WD, et al. Zhonghua Er Ke Za Zhi. *Chin J Pediatr.* (2021) 59:14–9.
17. Hu YM, Jiang ZF, Shen KL. *Zhufangtang Practical Pediatrics [M].* 8th ed. Beijing: People’s Medical Publishing House (2015). 1268–72.
18. Shen CF, Wang SM, Ho TS, Liu CC. Clinical features of community acquired adenovirus pneumonia during the 2011 community outbreak in Southern Taiwan: role of host immune response. *BMC Infect Dis.* (2017) 17:964–206. doi: 10.1186/s12879-017-2272-5
19. Buy NJ, Ghossain M. *Surface Epithelial-Stromal Tumors of the Ovary.* Berlin: Heidelberg: Springer-Verlag (2013).
20. Vargas HA, Micco M, Hong SI, Goldman DA, Dao F, Weigelt B, et al. Association between morphologic CT imaging traits and prognostically relevant gene signatures in women with high-grade serous ovarian cancer: a hypothesis-generating study. *Radiology.* (2015) 274:742–51. doi: 10.1148/radiol.14141477
21. van Nimwegen LWE, MarvinKuve-Groothuis AMC, de Krijger RR, Hulsuer CCC, Goverde AI, Zsirós J, et al. MR imaging in discriminating between benign and malignant pediatric ovarian masses: a systematic review. *Eur Radiol.* (2020) 30:1166–8. doi: 10.1007/s00330-019-06420-4
22. Guerriero S, Mallarini G, Ajossa S, Risalvato A, Satta R, Mais V, et al. Transvaginal ultrasound and computed tomography combined with clinical parameters and CA-125 determinations in the differential diagnosis of persistent ovarian cysts in premenopausal women. *Ultrasound Obstet Gynecol.* (1997) 9:339–43. doi: 10.1046/j.1460-0523.1997.0050339.x
23. Gemen O, Glađevich M, Ravid M, Piuza R, Rabinovich A, Gasper T, et al. A multicenter validation of computerized tomography models as predictors of non-optimal primary cytoreduction of advanced epithelial ovarian cancer. *Eur J Surg Oncol.* (2009) 35:1109–12. doi: 10.1016/j.ejso.2009.03.002
24. Yip SF, Liu Y, Parmar C, Li Q, Liu S, Qu F, et al. Associations between radiologist-defined semantic and automatically computed radiomic features in non-small cell lung cancer. *Sci Rep.* (2017) 7:3519. doi: 10.1038/s41598-017-02425-5

Conflict of Interest: HL was employed by the company GE Healthcare.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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