1.INTRODUCTION

1.1 Overview

Lung cancer survival rate is very limited post-surgery irrespective of if it is small cell or

non-small cell. A lot of work has been carried out by employing machine learning in life

expectancy prediction post thoracic surgery for patients with lung cancer. Many

machine learning models like multi-layer perceptron (MLP), SVM, naïve Bayes, decision

tree, random forest, logistic regression have been applied for post thoracic surgery life

expectancy prediction based on data sets from UCI. Also, work has been carried out

towards attribute ranking and selection in performing better in improving prediction

accuracy with machine learning algorithms. So accordingly, the authors, here, have

developed a deep neural network-based approach in prediction of post thoracic life

expectancy which is the most advanced form of neural networks. This is based on

dataset obtained from Wroclaw Thoracic Surgery Centre machine learning repository

which contained 470 instances. On comparing the accuracy, the results indicate that the

deep neural network can be efficiently used for predicting the life expectancy.

1.2 Purpose

In order to improve quality initiatives, healthcare administration, and consumer

education, it is critical to track health outcomes. The data obtained from patients who

had large lung resections for primary lung cancer is referred to as thoracic surgery.

Attribute ranking and selection are critical components of successful health outcome

prediction when using machine learning algorithms. Researchers used several

procedures, such as early-stage examinations, to determine the type of cancer before

symptoms appeared. The most relevant attributes are identified using attribute ranking

and selection, and the duplicated and unnecessary attributes are removed from the

dataset. The goal of our study is to look at patient mortality over the course of a year

after surgery. More precisely, we\'re looking into the patients\' underlying health issues,

which could be a powerful predictor of surgical-related mortality.

2.LITERATURE SURVEY

2.1 Existing Problem

It has always been a difficult task to accurately predict the life expectancy post an

operation. The prediction relies upon on numerous fitness elements of which a few have

a far important function in comparison to the opposite elements. A famous approach

used within side the beyond became to investigate the CT test photos of the lungs and

expect primarily based totally at the everyday check-ups. The thirty-day mortality price

is one statistic that has been used to estimate mortality charges within side the beyond.

This statistic, however, won't be absolutely correct due to the fact many sufferers die or

come to be very frail right now after this time period, requiring them to be transferred to

any other organization earlier than passing death. As a result, a large number of these

deaths go unreported.

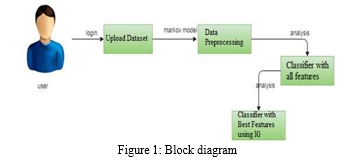
2.2 Proposed Solution

In order to improve quality initiatives, healthcare administration, and consumer education, it is critical to track health outcomes. Data from patients who have undergone extensive lung resection for primary lung cancer is called thoracic surgery. Attribute ranking and selection are important components for correctly predicting health outcomes when using machine learning algorithms. Researchers have used several techniques, such as early screening, to determine the type of cancer before symptoms appear. The most relevant attributes / characteristics are identified using attribute ranking and selection, and duplicate unwanted attributes / characteristics are removed from the dataset.

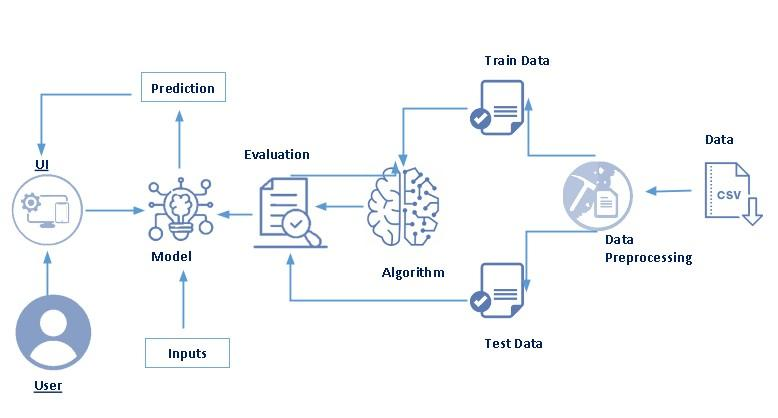
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3. Theoretical Analysis

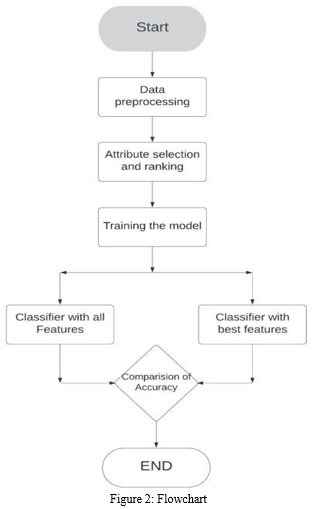
3.1 Block Diagram



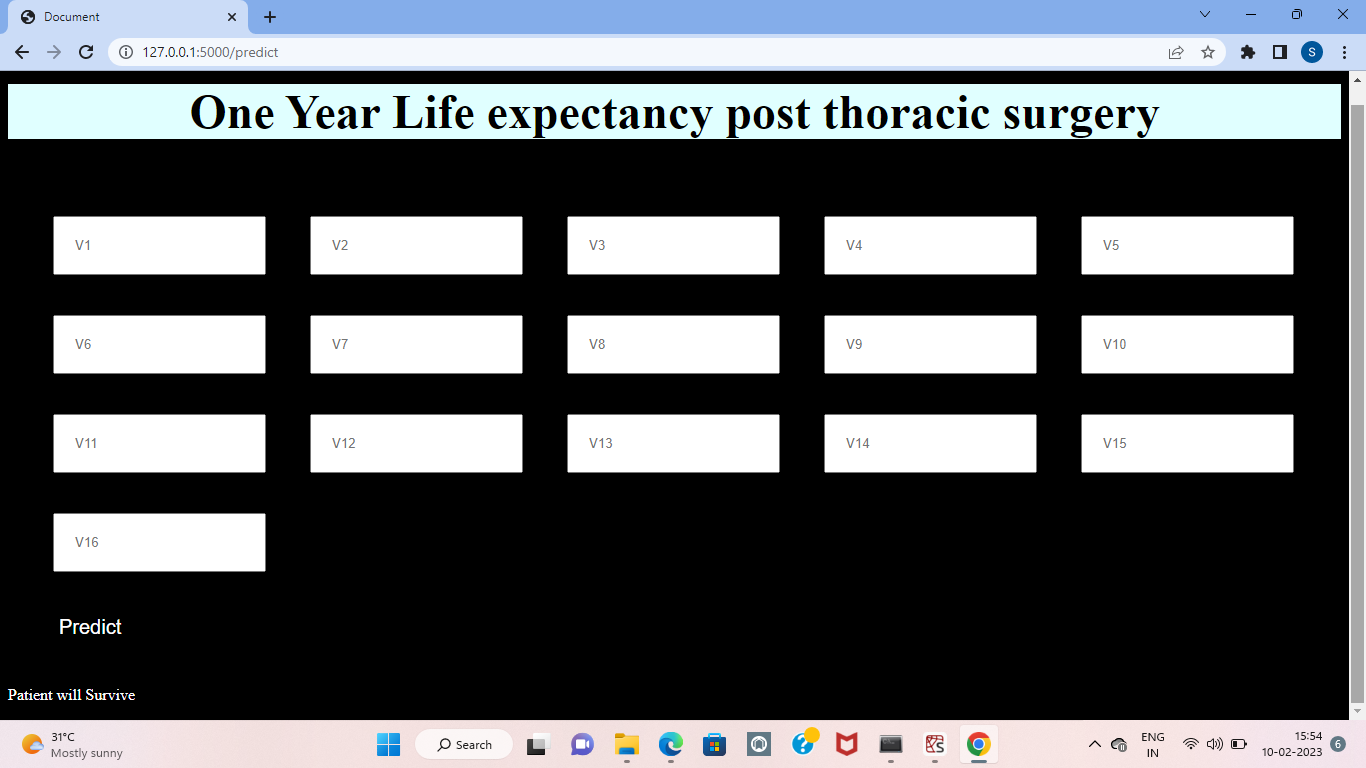
3.2 Hardware/Software designing



5. Flowchart



6. Result



7. Advantage and Disadvantage

Benefits includes enhancing predictability that facilitates the derivation of more

appropriate decisions. A study conducted by Chong et al. revealed this benefit since the

random forest classifier (RFC) ML model applied was cost-efficient ([19](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B19)). It also

prompted the prediction of lymph node metastasis during or after surgery for early-stage

adenocarcinomas with a sensitivity and specificity of 87.5 and 82.2 percent, respectively

([19](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B19)). In a similar perspective, Wu et al. indicated that the RFC machine learning

algorithm was the best predictive model and facilitated the recognition of lymph node

metastasis among patients with early T-stage non-small cell lung cancer before

undergoing an operation ([20](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B20)). Another study by Liu et al. revealed that machine learning

algorithms had the potential to assist in personalizing treatment decisions ([21](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B21)). This

conclusion arose from the excellent performance of the nomogram comprising CT-

derived radiomic features and risk factors developed by the authors in predicting post-

surgical progression-free survival of patients with lung adenocarcinoma ([21](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B21)).

Additionally, Liang et al. associated machine learning algorithms with improved lung

cancer screening and better assessment of treatment efficiency after the ML-assisted

deep methylation sequencing detected tumor-derived signals among patients with

surgery-resectable lung cancer ([22](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B22)). These arguments imply that the advantages

derived from application of ML in lung surgery ensure accurate predictability and

opportunities for enhancing patient outcomes.

On the contrary, the main shortcoming of machine learning in lung surgery is the lack of

sufficient research on using them appropriately. For instance, a study by Feng et al.

revealed that most ML techniques require adopting different protocols and settings that

require complex analysis in literature to avoid any errors ([23](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B23)). The second limitation of

ML algorithms in lung surgery is the tendency of most outcome prediction researchers

to adopt a small dataset that increases the chances of biased study results ([24](https://www.ncbi.nlm.nih.gov/pmc/articles/PMC9510630/#B24)).

Therefore, it remains vital to deal with the general challenges facing ML implementation

to ensure the successful integration of the concept into lung surgery. This factor

affected the derived results since the researchers could have excluded articles in other

languages that provided information relevant to the study. The second limitation of the

study is a wide literature gap in the field. This conclusion arises from the derivation of

only nineteen articles to include in the research. The final limitation stems from adopting

the narrative review approach since the technique led to broad findings, making it hard

to draw essential conclusions.

8. Applications

* Rapid detection and screening of post-surgical complications
* Rapidly diagnose or screen cardiac abnormalities
* Objective and consistent aortic diameter measurement
* Objectively and rapidly measure cardiac function in real time
* Screening high-risk populations and basic research
* Stratifying patient with type A aortic dissection for closer monitoring and follow-up
* Stratifying patients for closer monitoring and follow-up
* Rapid and continuous surgical skill evaluation without need for manual annotation
* Real-time technique feedback in the operating room
* Inform surgical risks of high-risk patients
* Risk stratify post-operative patients for closer follow-up
* Increase the scale and speed of basic research projects by eliminating manual diameter measurement

9.CONCLUSION

In this study, the quality of the three classification methods/ algorithms have been tested

to improve the prognosis for the life of patients with thoracic thoracic cancer after

thoracic cancer surgery. Three ways to train the machine before and after using the

quality level options are compared with their improved versions. The results show that

boosting is not always the best solution, where the level of responsibility and choice can

make better at improving predictive accuracy. Other qualifications and machine learning

strategies can be introduced in the future work to find the best performance of the data

forecast model. The results indicate that decision tree, logistic regression and KNN give

us better performance than other data mining algorithms, we also have monitored and

mentioned the performance of each algorithm. From this study we deducted that FEV1,

Dyspnoea before surgery, Size of the tumour, Weakness before surgery and

performance status are the attributes/features that are mainly responsible for the

survival of patients after the thoracic surgery. This study asserts the use of data mining

algorithms in medical field as the results are better than the existing system which are

confirmed by statistical analysis.

10.FUTURE SCOPE

This study does have few limitations. The results/outputs which were obtained are particular to a nation or an organization which collected the data set. Results obtained may be time-limited (2007–11). The dataset used as part of this project has very less records and may impede the accuracy of few algorithms that are used. In any case, this dataset can serve as a starting point to raise a better understanding of thoracic surgery patients. These tests can be further extended.

This analysis only makes and uses three data mining methods. Therefore, some more machine learning methods could be used to get more knowledge about the data-set as a future work.

11.BIBLIOGRAPHY

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