

SPOTTING LUNG AND COLON CANCER USING HYBRID ENSEMBLE LEARNING

Under the Supervision of

Mrs.K.Thrilochana Devi *M. Tech*

Presented By:

1.A.Tharun	19BQ1A1208
2.B.Ajay	19BQ1A1214
3.B. Vasu	19BQ1A1215
4.CH.Teja krishna Reddy	19BQ1A1221

Contents

1. Proposed System with Advantages
2. Software Requirements
3. Hardware Requirements
4. Functional Requirements and Non-Functional requirements
5. System Architecture
6. System Design
7. Output Screens

Proposed system with Advantages

- In this proposed model we use a hybrid approach which is a combination of CNN model and Optimizer techniques.
- We have covered image pre-processing, deep feature extraction, optimizer, confusion matrix
- We also give a basic CNN algorithm with an Optimizer that are used to detect lung and colon cancer efficiently.
- Proposed model using feature extraction and CNN algorithm on LC25000 lung and colon histopathological image datasets to assure a prognosis of lung and colon cancer.
- Image classification and Segmentation analyze.
- Less amount of Time and low budget to recovery of patients.
- Histopathology is main role in Fast performance.

Software Requirements

- Operating System : Microsoft Windows 08 or above
- Code Editor : Google colab
- Languages : Python 3.7 or above
- Libraries : Pandas, Numpy, Matplotlib, Keras, Scikit-learn

Hardware Requirements

- Hard disk : 1TB HDD
- Ram : 4GB or above
- Processor : i3 or above

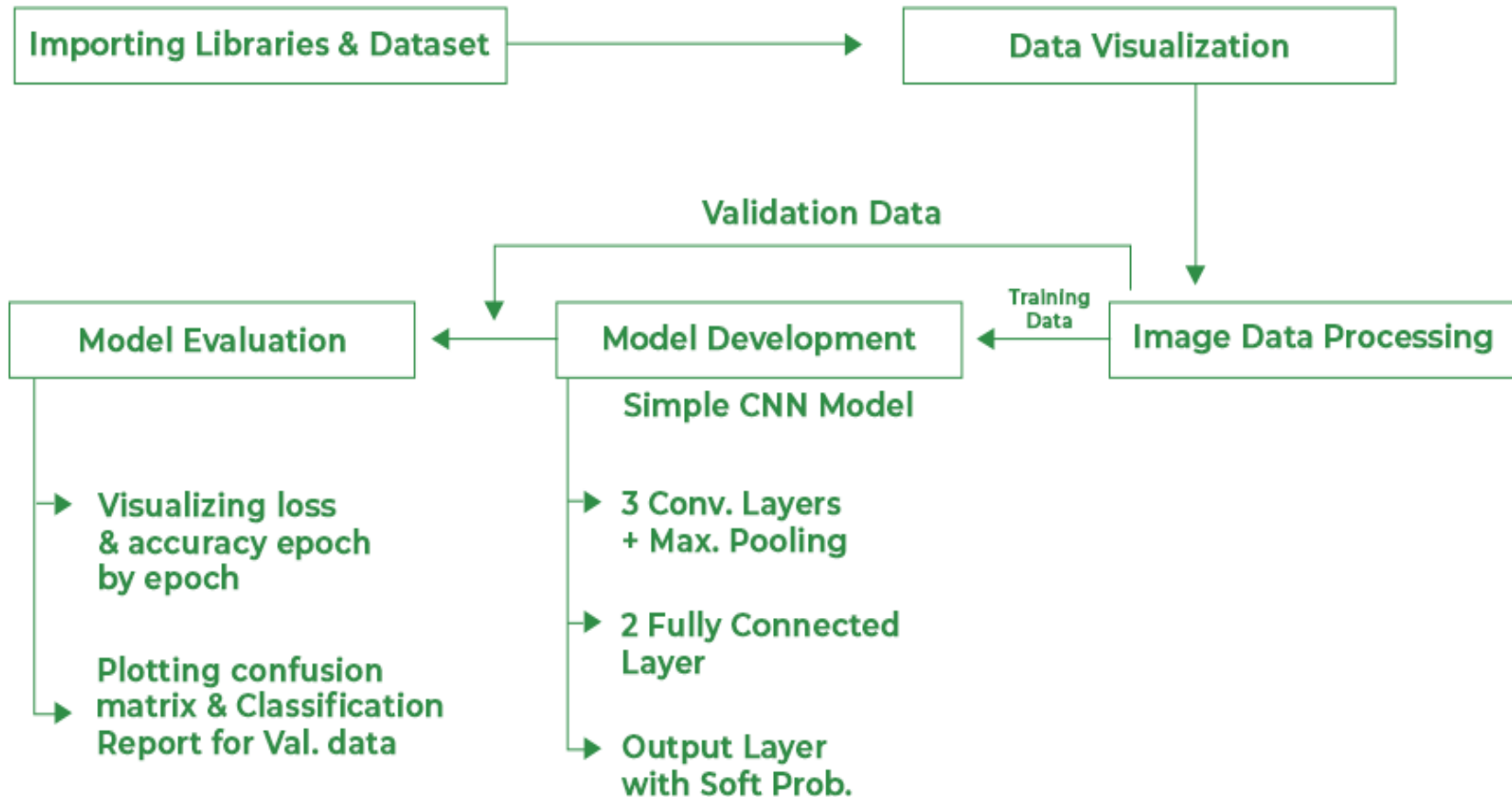
Functional Requirements

- **Data preprocessing:** The ability to clean and prepare the data for training and testing the models.
- **Feature extraction:** The ability to extract relevant features from the data that can be used to train the models.
- **Model training:** The ability to train the selected models on the processed and feature-extracted data using CNN models such as VGG16,VGG19.
- **Model deployment:** The ability to deploy the trained models in a clinical setting for real-time prediction of lung cancer.
- **Model monitoring :** The ability to monitor the performance of the deployed models over time and update them as needed.

Non-functional requirements

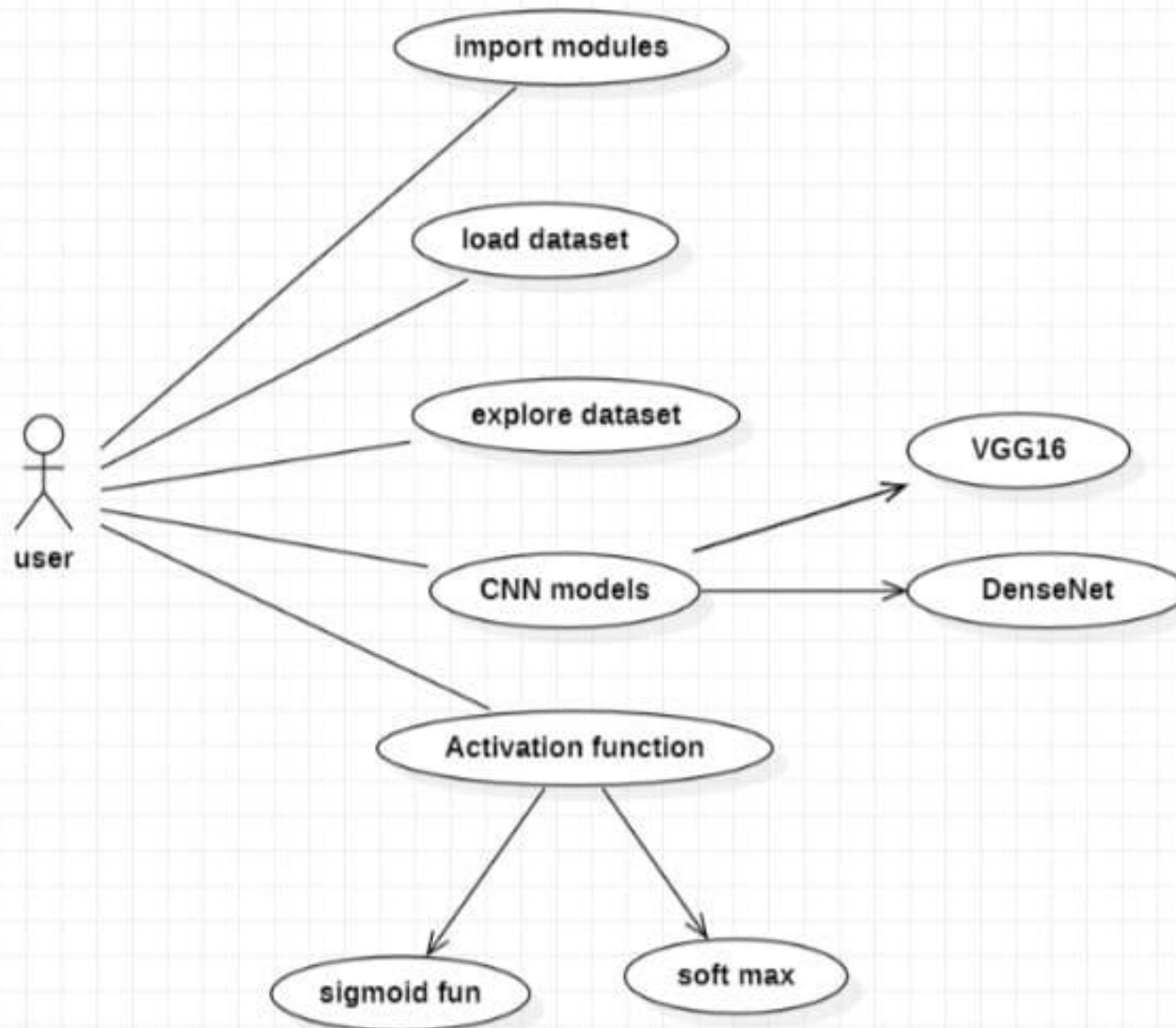
- **Performance:** The system should have a fast processing time for real-time predictions and a high accuracy rate for detecting lung cancer.
- **Scalability:** The system should be able to handle a large amount of data and be able to adapt to changing data sizes and types.
- **Usability:** The system should be user-friendly for radiologists and medical professionals, and provide clear and easy-to-understand results.
- **Security:** The system should be able to protect patient data and prevent unauthorized access.
- **Maintainability:** The system should be easy to maintain and update over time.
- **Availability:** The system should be available and accessible to users when needed, with minimal downtime.

System Architecture

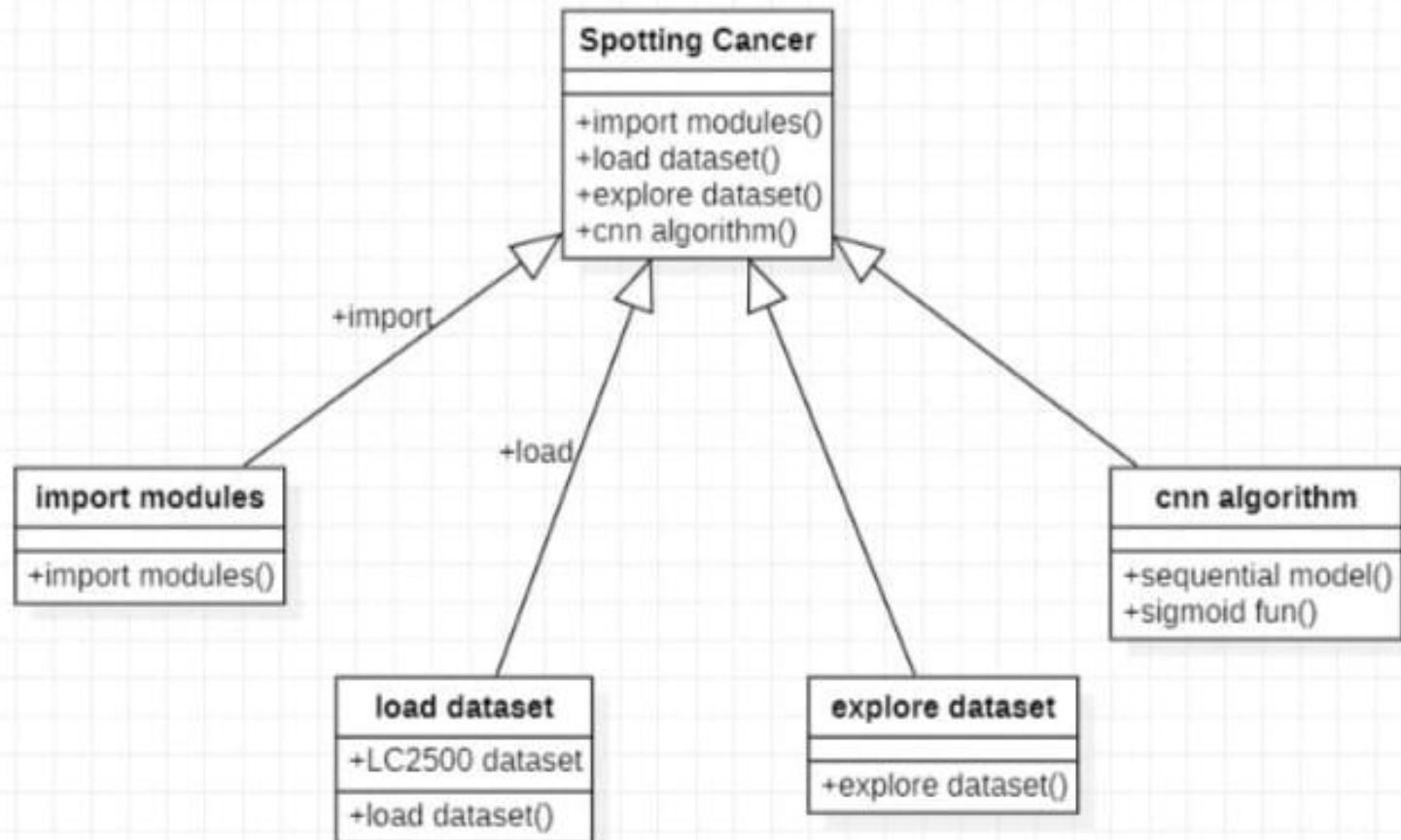


System Design

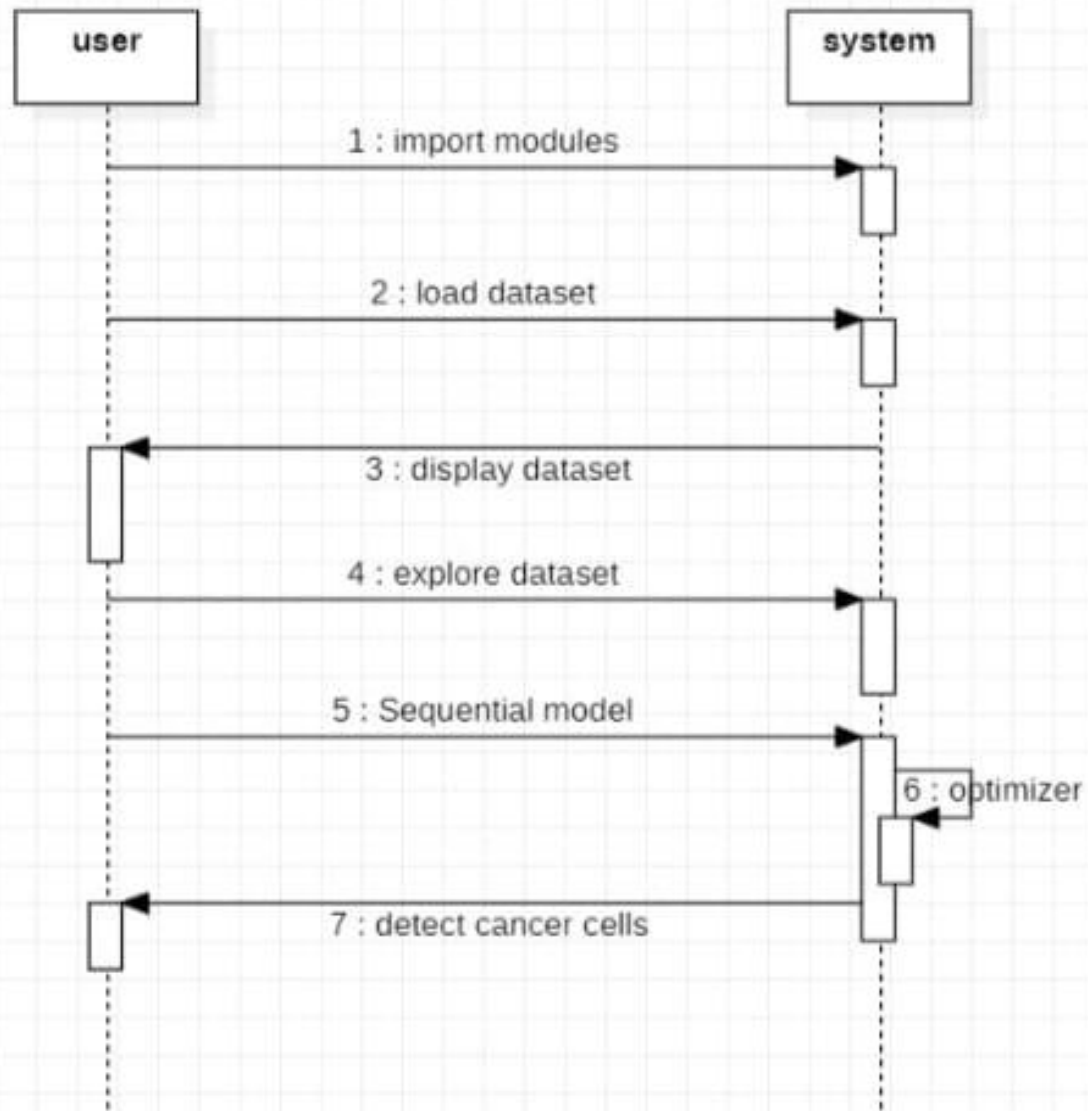
- **Use Case Diagram** :Use case diagram is a diagram showing a series of use cases and actors of use and their relation.
- Graphically it is made as a solid line ellipse, with only its name included.
- The main actors of Recognition of Sign Language System are:
 - 1) User
 - 2) System
 - 3) Web Cam



- **Class Diagram:** Class diagrams model class structure and contents using design elements such as classes, packages and objects.
- Classes are composed of three things: name, attributes, and operations.
- Class Diagram association relationship is most common relationship in a class diagram.
- The association shows the relationship between instances of class.

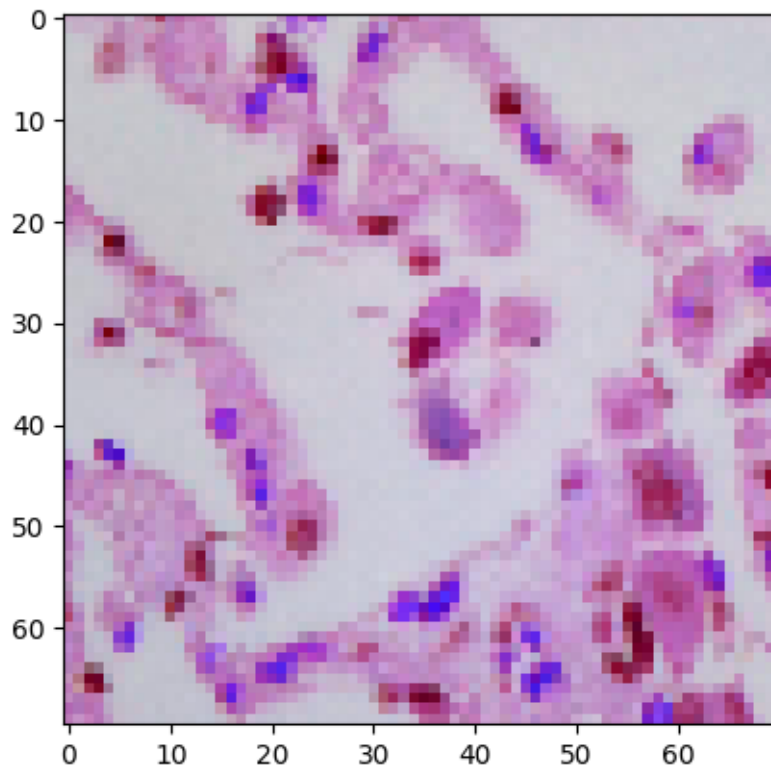


- **Sequence Diagram:** Sequence diagrams also called as INTERACTION DIAGRAMS.
- An interaction diagram represents an interaction, which consists of a series of objects and their relationship and the messages that can be exchanged between them.
- A sequence diagram empathizes the time ordering of messages.
- Graphically a sequence diagram is a table that shows objects arranged along the X-axis and messages ordered in increasing time along the Y-axis.



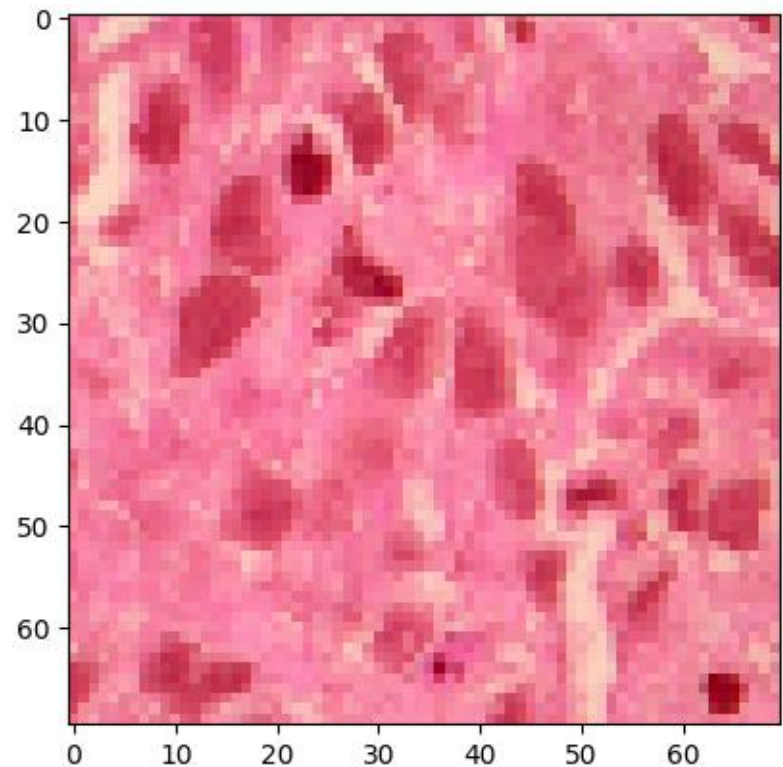
Output Screens

Before :



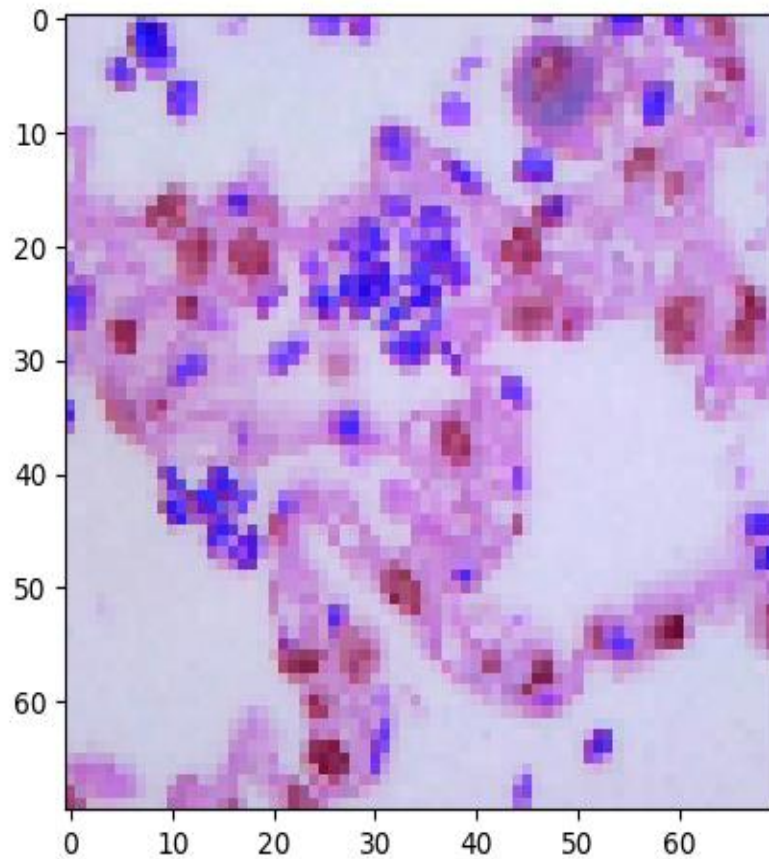
Predict val: lung_scc

After :



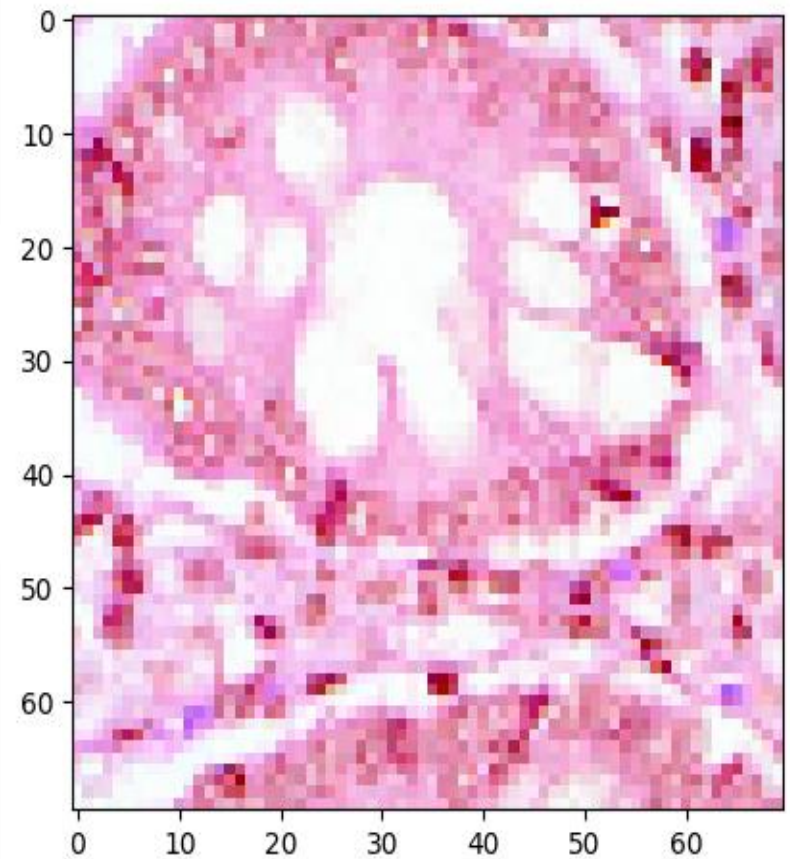
Real val : lung_scc

Before :



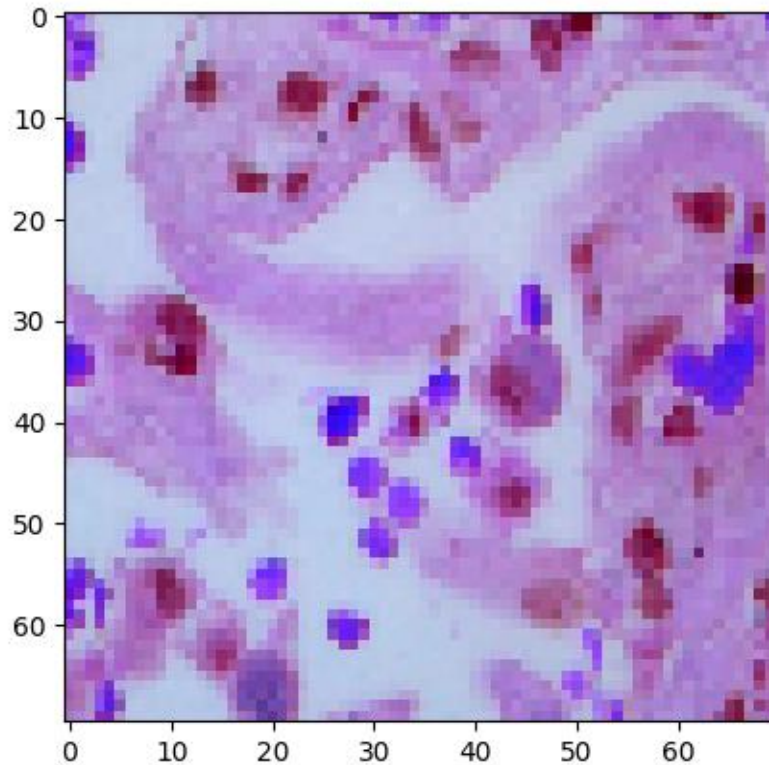
Predict val : Colon_aca

After:



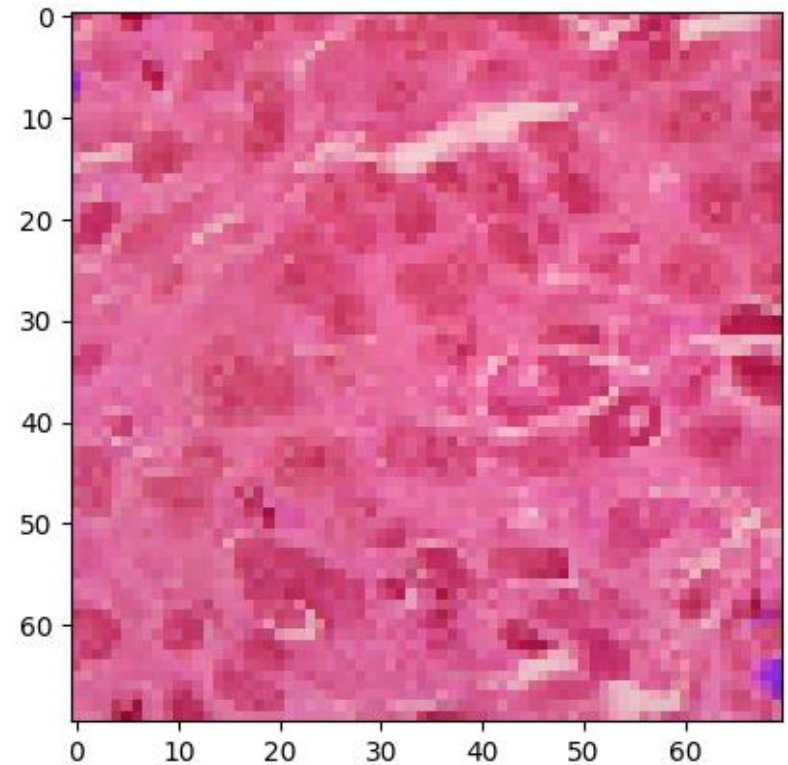
Real val : colon_aca

Before :



Predict val : lung_aca

After :



Real val : lung_aca



Thank you!