Problem Description and Data Overview

Git: https://github.com/KQian-lab/kagglehw5

Problem:

The task of this competition was to classify histopathological images of lymph node sections into either cancerous or non-cancerous categories.

Data Overview

Training labels: Contains image IDs and the corresponding labels indicating the presence of cancer, 1, or absence, 0.

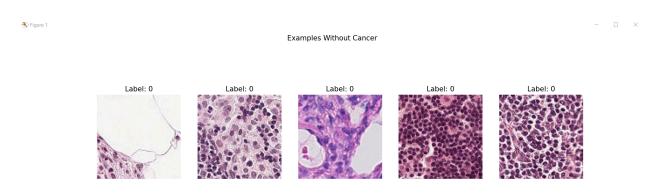
Dimensions: Images are processed into 96 x 96 pixels with three color channels.

Data Size

- 220,025 Training Images
- 57,458 Test Images

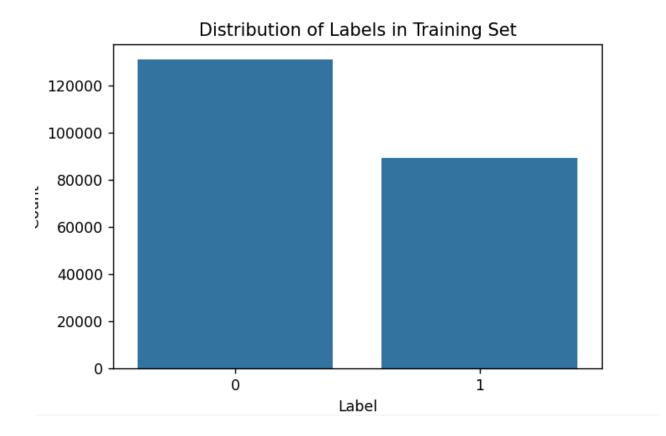
Exploratory Data Analysis and Data Cleaning

The data contained no missing values and image dimensions were ensured to remain consistent



 ${f \&}$ Figure 1 - $f \Box$ imes Examples With Cancer





Model Architecture

The model used is a Convolutional Neural Network for binary classification.

Convolutional Neural Networks are good at processing images because they can automatically

detect patterns and features, such as edges and textures, which are important for classifying

images. The output layer uses a sigmoid activation function which is better for binary

classification tasks. In this case, identifying if an image is cancerous or not.

Results and Analysis

Training and Validation:

• Epochs: 6

Batch Size: 32

Performance measurement: Accuracy and AUC

Issues Encountered:

• System Resources - Lack of GPU processing resulted in very long processing times for

each epoch

o This was resolved by using a much smaller random sample of images instead of

the entire set

Hyperparameters:

Learning Rate: Set by the Adam optimizer, 0.001

• Batch Size: 32

Number of Filters in Convolutional Layers: 32, 64, 128

• Kernel Size: (3,3)

• Dropout Rate: 0.25 for convolutional layers, 0.5 in dense layers

Number of Neurons in Dense Layer: 256

Activation Functions: ReLU

Epochs: 6

Submission and Description

Private Score ①

Public Score ①

Submission1.csv
Complete (after deadline) · 15s ago

O.7146

Conclusion

This implemented CNN model is effective for processing histopathological images as cancerous or non-cancerous. I believe if it were possible to train the model on the entire dataset instead of a small random sample, the results would have been much better.