

Classification prediction of non-small cell lung cancer

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Introduction

Nowadays, lung cancer is a leading cause of death in both men and women worldwide.[1] Lung Adenocarcinoma (LUAD) and Lung Squamous Cell Carcinoma (LUSC) are the most prevalent subtypes of lung cancer, and their distinction requires visual inspection by an experienced pathologist. [2] Nowadays, more and more researchers use machine learning to solve this question, such as convolutional neural network. In deep learning, a convolutional neural network (CNN) is a class of deep neural networks, most commonly applied to analyzing visual imagery.[3]

Dataset used

We obtain whole-slide images of LUAD and LUSC from the Genomic Data Commons (GDC) portal which were generated by The Cancer Genome Atlas (TCGA), and they have made them publicly available. [4] The GDC is a research program of the National Cancer Institute (NCI). The mission of the GDC is to provide the cancer research community with a unified data repository that enables data sharing across cancer genomic studies in support of precision medicine. [5]

Figure 1a shows an example of slide images and Figure 1b is the basic information. For later analysis, we classify the dataset into three groups: normal tissue, LUAD and LUSC. The number of corresponding cases is respectively 528, 585 and 504. And the slides with a low information will be removed.

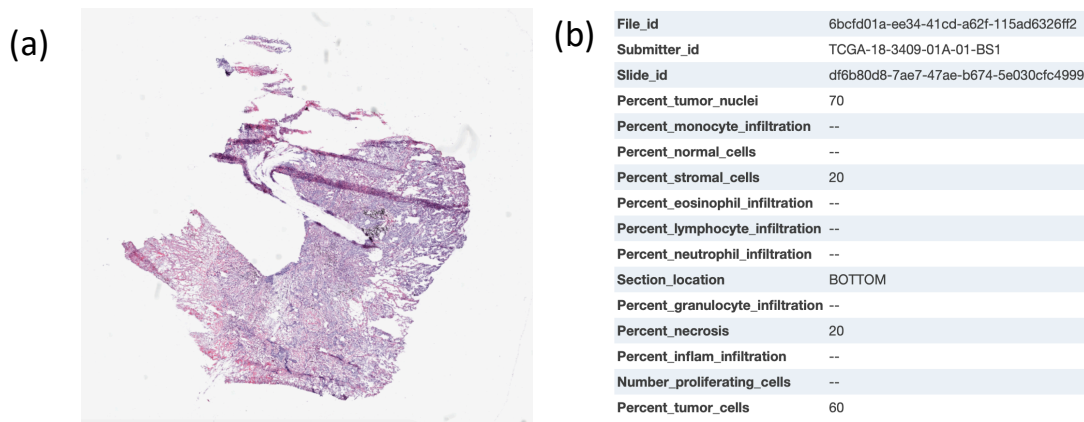


Figure 1. Dataset details. (a) Eosin stained histopathology images. (b) The details of the left picture.

Goal & Question

Use convolutional neural network or other deep-learning models to assist pathologists in the detection of lung cancer subtype. And, how to optimize the model to make the prediction more accurate?

References

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