lung-cancer-efficientnet-4

July 14, 2024

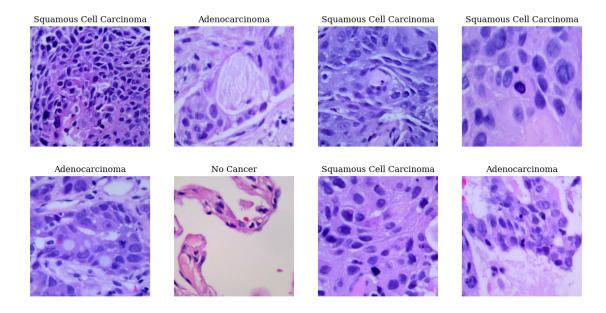
[1]: import os

```
import matplotlib.pyplot as plt
     import numpy as np
     import pandas as pd
     import tensorflow as tf
     from tensorflow.keras import layers
     import seaborn as sns
     from sklearn.metrics import confusion_matrix, classification_report
     import cv2
    2024-04-01 15:31:04.337542: E
    external/local xla/xla/stream_executor/cuda/cuda_dnn.cc:9261] Unable to register
    cuDNN factory: Attempting to register factory for plugin cuDNN when one has
    already been registered
    2024-04-01 15:31:04.337651: E
    external/local_xla/xtream_executor/cuda/cuda_fft.cc:607] Unable to register
    cuFFT factory: Attempting to register factory for plugin cuFFT when one has
    already been registered
    2024-04-01 15:31:04.494395: E
    external/local xla/xla/stream executor/cuda/cuda blas.cc:1515] Unable to
    register cuBLAS factory: Attempting to register factory for plugin cuBLAS when
    one has already been registered
[2]: SEED = 15243
     np.random.seed(SEED)
     os.environ["PYTHONHASHSEED"] = str(SEED)
     tf.random.set_seed(SEED)
     plt.rc("axes.spines", right=False, top=False)
     plt.rc("font", family="serif")
[3]: BATCH SIZE = 25
     DATA_DIR = ("/kaggle/input/lung-and-colon-cancer-histopathological-images/
      →lung_colon_image_set/lung_image_sets")
     IMG_SIZE = 256
     MAX EPOCHS = 100
     class_names = ["lung_n", "lung_aca", "lung_scc"]
```

```
class_details = ["No Cancer", "Adenocarcinoma", "Squamous Cell Carcinoma"]
[4]: train_ds = tf.keras.preprocessing.image_dataset_from_directory(
         DATA DIR,
         batch_size=BATCH_SIZE,
         image_size=(IMG_SIZE, IMG_SIZE),
         class_names=class_names,
         seed=45,
         subset="training",
         validation_split=0.2
     )
    Found 15000 files belonging to 3 classes.
    Using 12000 files for training.
[5]: validation_ds = tf.keras.preprocessing.image_dataset_from_directory(
         DATA_DIR,
         batch_size=BATCH_SIZE,
         image_size=(IMG_SIZE, IMG_SIZE),
         class_names=class_names,
         seed=45,
         subset="validation",
         validation_split=0.2
    Found 15000 files belonging to 3 classes.
    Using 3000 files for validation.
[6]: plt.figure(figsize=(14, 7))
     for images, labels in train_ds.take(1):
         labels = labels.numpy()
         for i in range(8):
             ax = plt.subplot(2, 4, i + 1)
             plt.imshow(images[i].numpy().astype("uint8"))
```

plt.title(class_details[labels[i]])

plt.axis("off")



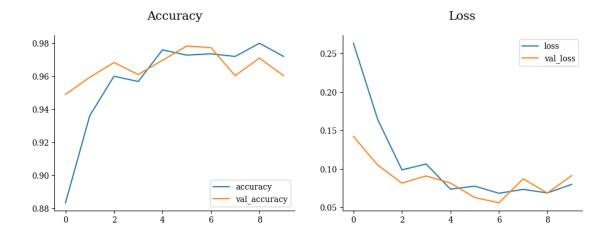
```
[7]: AUTOTUNE = tf.data.AUTOTUNE
     train_ds = train_ds.cache(".cached-data").prefetch(buffer_size=AUTOTUNE)
     validation_ds = validation_ds.cache().prefetch(buffer_size=AUTOTUNE)
     def compile_and_fit_model(model: tf.keras.Sequential) -> tf.keras.callbacks.
      →History:
         model.compile(
             optimizer="adam",
             loss="sparse_categorical_crossentropy",
             metrics=["accuracy"]
         early_stopping = tf.keras.callbacks.EarlyStopping(
             monitor="val_loss",
             min_delta=1e-4,
             patience=5,
             mode="auto",
             restore_best_weights=True
         reduce_lr = tf.keras.callbacks.ReduceLROnPlateau(
             monitor="val_loss",
             factor=0.2,
             patience=5,
             min_lr=0.001
         history = model.fit(
             train_ds,
             validation_data=validation_ds,
             epochs=MAX_EPOCHS,
             steps_per_epoch=50,
```

```
callbacks=[early_stopping, reduce_lr]
        )
        performance_df = pd.DataFrame(history.history)
        fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(12, 4))
        for ax, metric in zip(axes.flat, ["accuracy", "loss"]):
           performance_df.filter(like=metric).plot(ax=ax)
            ax.set_title(metric.title(), size=15, pad=20)
        return history
[8]: pretrained_efficientnet_base = tf.keras.applications.efficientnet.
     →EfficientNetB1(
        include_top=False, weights="imagenet", pooling="avg",
    )
    pretrained_efficientnet_base.trainable = False
    efficientnet_model = tf.keras.Sequential([
        layers.Input(shape=(IMG_SIZE, IMG_SIZE, 3)),
        pretrained_efficientnet_base,
        layers.Dense(128, activation="relu"),
        layers.Dense(3, activation="softmax")
    ])
    efficientnet_model.summary()
   Downloading data from https://storage.googleapis.com/keras-
   applications/efficientnetb1_notop.h5
   Model: "sequential"
    Layer (type)
                              Output Shape
                                                      Param #
    ______
    efficientnetb1 (Functional (None, 1280)
                                                      6575239
    )
    dense (Dense)
                              (None, 128)
                                                      163968
    dense_1 (Dense)
                              (None, 3)
                                                      387
   Total params: 6739594 (25.71 MB)
   Trainable params: 164355 (642.01 KB)
   Non-trainable params: 6575239 (25.08 MB)
[9]: efficientnet_history = compile_and_fit_model(efficientnet_model)
   Epoch 1/100
```

tensorflow/core/grappler/optimizers/meta_optimizer.cc:961] layout failed:

2024-04-01 15:31:51.657213: E

```
INVALID_ARGUMENT: Size of values 0 does not match size of permutation 4 @ fanin
shape
insequential/efficientnetb1/block1b_drop/dropout/SelectV2-2-TransposeNHWCToNCHW-
LayoutOptimizer
WARNING: All log messages before absl::InitializeLog() is called are written to
STDERR
I0000 00:00:1711985515.543982
                          109 device compiler.h:186] Compiled cluster
using XLA! This line is logged at most once for the lifetime of the process.
50/50 [============= ] - 33s 394ms/step - loss: 0.2633 -
accuracy: 0.8832 - val_loss: 0.1419 - val_accuracy: 0.9490 - lr: 0.0010
Epoch 2/100
accuracy: 0.9360 - val_loss: 0.1049 - val_accuracy: 0.9593 - lr: 0.0010
Epoch 3/100
accuracy: 0.9600 - val_loss: 0.0813 - val_accuracy: 0.9683 - lr: 0.0010
Epoch 4/100
accuracy: 0.9568 - val_loss: 0.0907 - val_accuracy: 0.9610 - lr: 0.0010
Epoch 5/100
50/50 [============ ] - 13s 264ms/step - loss: 0.0734 -
accuracy: 0.9760 - val_loss: 0.0814 - val_accuracy: 0.9697 - lr: 0.0010
Epoch 6/100
50/50 [============ ] - 13s 266ms/step - loss: 0.0774 -
accuracy: 0.9728 - val_loss: 0.0626 - val_accuracy: 0.9783 - lr: 0.0010
Epoch 7/100
accuracy: 0.9736 - val_loss: 0.0557 - val_accuracy: 0.9773 - lr: 0.0010
Epoch 8/100
accuracy: 0.9720 - val_loss: 0.0869 - val_accuracy: 0.9603 - lr: 0.0010
Epoch 9/100
50/50 [=========== ] - 13s 264ms/step - loss: 0.0685 -
accuracy: 0.9800 - val_loss: 0.0684 - val_accuracy: 0.9710 - lr: 0.0010
Epoch 10/100
50/50 [============ ] - 12s 238ms/step - loss: 0.0797 -
accuracy: 0.9720 - val_loss: 0.0911 - val_accuracy: 0.9603 - lr: 0.0010
```



```
[10]: '''def get_predictions(model, dataset):
          y_true = []
          y_pred = []
          for images, labels in dataset:
               y_true.extend(labels.numpy())
              predictions = model.predict(images)
               y_pred.extend(np.argmax(predictions, axis=1))
          return np.array(y_true), np.array(y_pred)
      train_true, train_pred = get_predictions(efficientnet_model, train_ds)
      val_true, val_pred = get_predictions(efficientnet_model, validation_ds)
      train_conf_matrix = confusion_matrix(train_true, train_pred)
      train\_class\_report = classification\_report(train\_true, train\_pred, 
       \hookrightarrow target\_names = class\_details)
      val_conf_matrix = confusion_matrix(val_true, val_pred)
      val_class_report = classification_report(val_true, val_pred,_
       \hookrightarrow target\_names = class\_details)
      print("Confusion Matrix (Training Set):\n", train_conf_matrix)
      print("\nClassification Report (Training Set):\n", train_class_report)
      print("\nConfusion\ Matrix\ (Validation\ Set):\n",\ val\_conf\_matrix)
      print("\nClassification\ Report\ (Validation\ Set):\n",\ val\_class\_report)'''
```

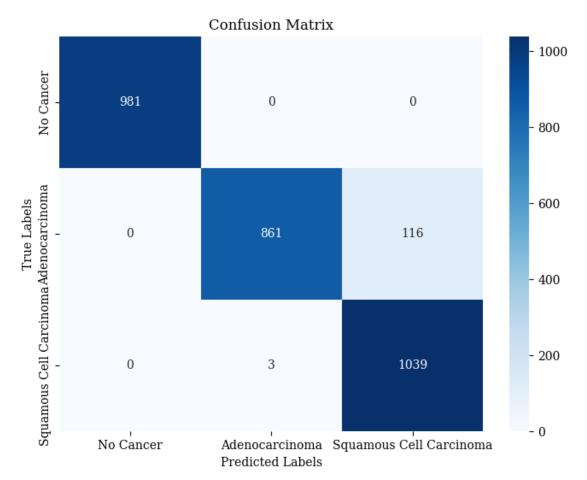
[10]: 'def get_predictions(model, dataset):\n y_true = []\n y_pred = []\n\n
for images, labels in dataset:\n y_true.extend(labels.numpy())\n

```
predictions = model.predict(images)\n
     y_pred.extend(np.argmax(predictions, axis=1))\n\n
                                                     return np.array(y_true),
     np.array(y_pred)\n\ntrain_true, train_pred =
     get_predictions(efficientnet_model, train_ds)\nval_true, val_pred =
     get_predictions(efficientnet_model, validation_ds)\n\ntrain_conf_matrix =
     confusion_matrix(train_true, train_pred)\ntrain_class_report =
     classification_report(train_true, train_pred,
     target_names=class_details)\n\nval_conf_matrix = confusion_matrix(val_true,
     val pred)\nval class report = classification report(val true, val pred,
     target_names=class_details)\n\nprint("Confusion Matrix (Training Set):\n",
     train_conf_matrix)\nprint("\nClassification Report (Training Set):\n",
     train_class_report)\n\nprint("\nConfusion Matrix (Validation Set):\n",
     val_conf_matrix)\nprint("\nClassification Report (Validation Set):\n",
     val_class_report)'
[11]: validation_predictions = []
     validation_labels = []
     for images, labels in validation_ds:
         validation_predictions.extend(np.argmax(efficientnet_model.predict(images),_
      →axis=1))
         validation_labels.extend(labels.numpy())
     conf_matrix = confusion_matrix(validation_labels, validation_predictions)
     class_report = classification_report(validation_labels, validation_predictions,_
      starget_names=class_details)
     plt.figure(figsize=(8, 6))
     sns.heatmap(conf_matrix, annot=True, fmt="d", cmap="Blues", __
      Axticklabels=class_details, yticklabels=class_details)
     plt.xlabel("Predicted Labels")
     plt.ylabel("True Labels")
     plt.title("Confusion Matrix")
     plt.show()
     print("Classification Report:")
     print(class_report)
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    1/1 [=======] - 0s 50ms/step
    1/1 [=======] - Os 39ms/step
    1/1 [======= ] - Os 49ms/step
    1/1 [=======] - Os 54ms/step
    1/1 [=======] - Os 54ms/step
     1/1 [====== ] - Os 41ms/step
     1/1 [======= ] - Os 50ms/step
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```

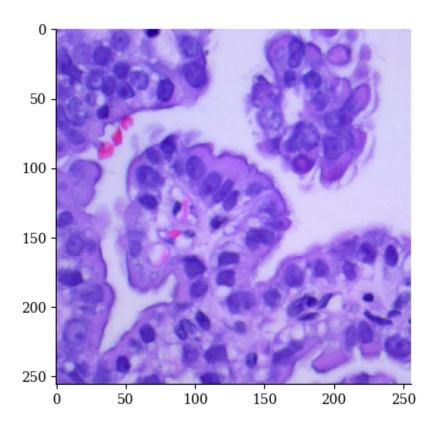


Classification Report:

precision recall f1-score support

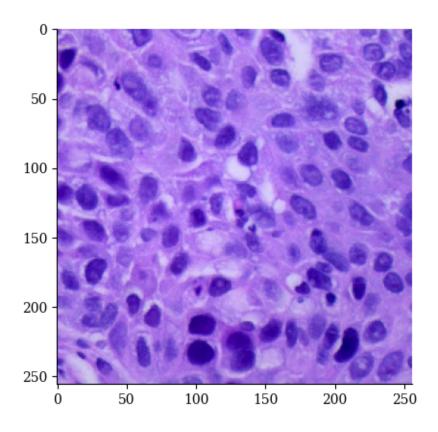
```
No Cancer
                              1.00
                                        1.00
                                                  1.00
                                                             981
         Adenocarcinoma
                              1.00
                                        0.88
                                                  0.94
                                                             977
Squamous Cell Carcinoma
                              0.90
                                        1.00
                                                  0.95
                                                             1042
                                                  0.96
                                                             3000
               accuracy
                                                  0.96
                                                             3000
              macro avg
                              0.97
                                        0.96
           weighted avg
                              0.96
                                        0.96
                                                  0.96
                                                             3000
```

```
[12]: prediction_folder="/kaggle/input/prediction-set/pred"
      image_files = os.listdir(prediction_folder)
      from keras.preprocessing import image
      model=efficientnet_model
      for image_file in image_files:
          print(image_file)
          img = os.path.join(prediction_folder, image_file)
          img = image.load_img(img,target_size=(IMG_SIZE, IMG_SIZE))
          imag = image.img_to_array(img)
          imaga = np.expand_dims(imag,axis=0)
          ypred = model.predict(imaga)
          print(ypred)
          a=np.argmax(ypred,-1)
          if a==0:
                op="no cancer "
          elif a==1:
                op="adinocarcinoma"
          else:
                op="squamous cell carcinoma"
          plt.imshow(img)
          plt.show()
          print("THE UPLOADED IMAGE IS SUSPECTED AS: "+str(op))
```



THE UPLOADED IMAGE IS SUSPECTED AS: adinocarcinoma lungscc1007.jpeg
1/1 [=======] - Os 30ms/step

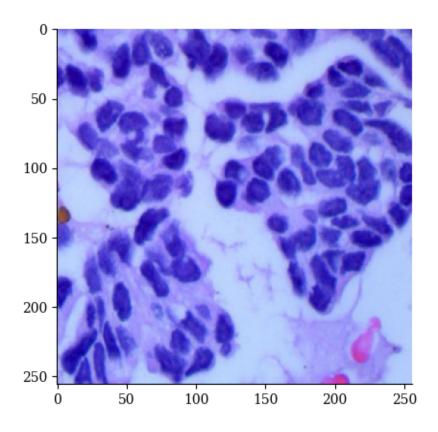
[[2.0589576e-08 8.1440201e-04 9.9918562e-01]]



THE UPLOADED IMAGE IS SUSPECTED AS: squamous cell carcinoma lungaca1013.jpeg $\,$

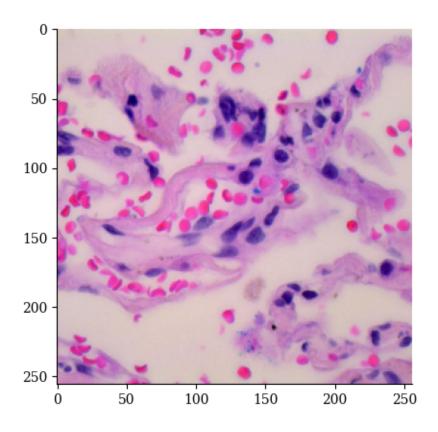
1/1 [======] - Os 32ms/step

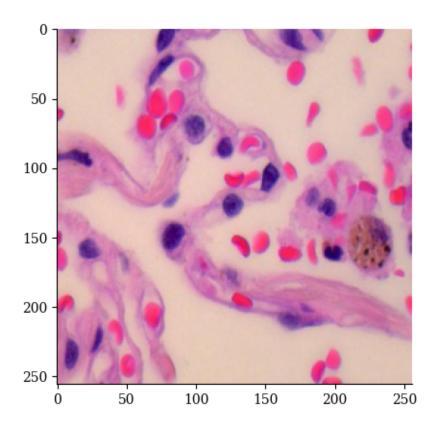
[[7.572352e-04 9.987735e-01 4.692499e-04]]

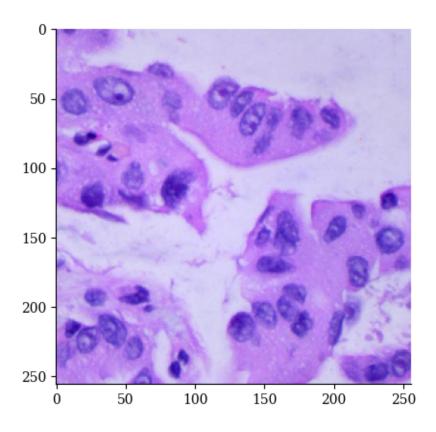


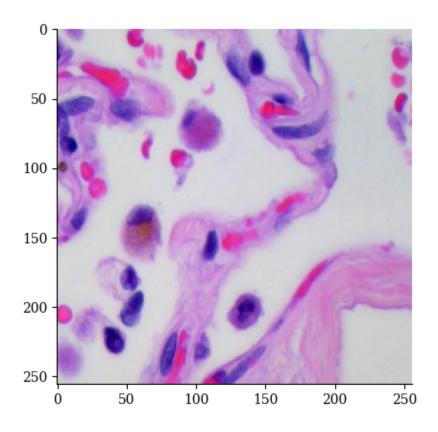
THE UPLOADED IMAGE IS SUSPECTED AS: adinocarcinoma lungn1009.jpeg

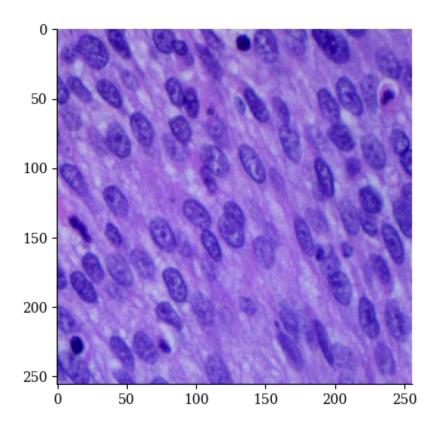
1/1 [=======] - 0s 30ms/step [[9.9999678e-01 3.1839709e-06 9.6623598e-09]]







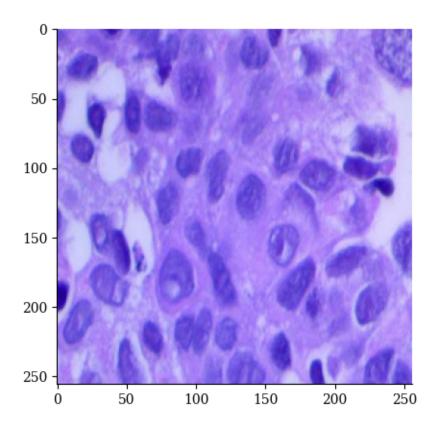




THE UPLOADED IMAGE IS SUSPECTED AS: squamous cell carcinoma lungscc1010.jpeg $\,$

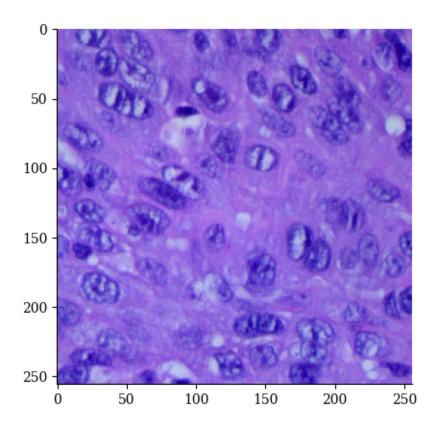
1/1 [======] - Os 31ms/step

[[4.0803428e-08 2.5102727e-02 9.7489727e-01]]



THE UPLOADED IMAGE IS SUSPECTED AS: squamous cell carcinoma lungscc1.jpeg

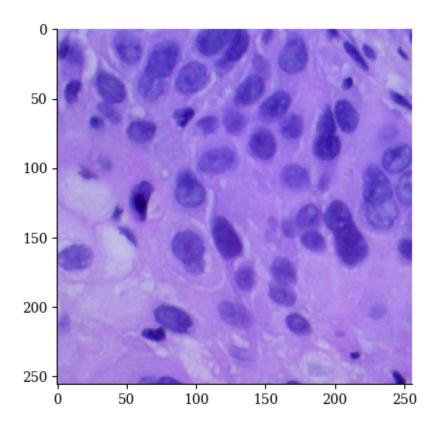
[[2.4777516e-09 6.8083033e-04 9.9931920e-01]]

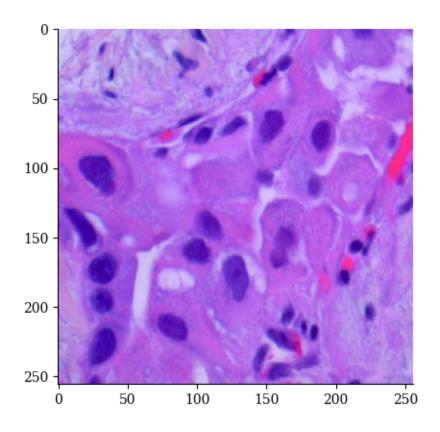


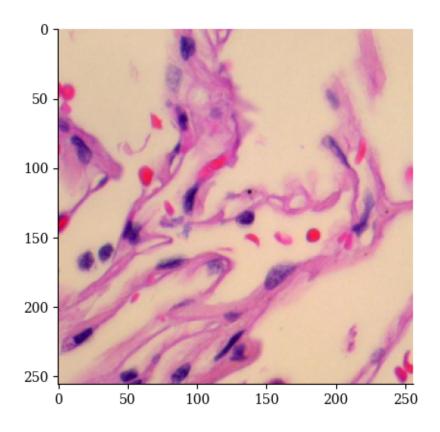
THE UPLOADED IMAGE IS SUSPECTED AS: squamous cell carcinoma lungscc1005.jpeg $\,$

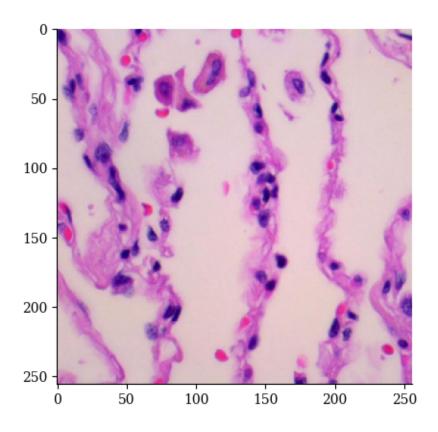
1/1 [======] - Os 31ms/step

[[2.2098314e-08 3.5357254e-04 9.9964643e-01]]





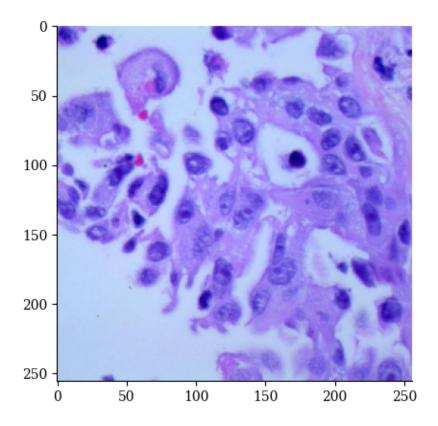




THE UPLOADED IMAGE IS SUSPECTED AS: no cancer lungaca1023.jpeg

1/1 [======] - Os 30ms/step

[[1.4573052e-05 9.9216568e-01 7.8197625e-03]]



THE UPLOADED IMAGE IS SUSPECTED AS: adinocarcinoma