

ML/DL Study

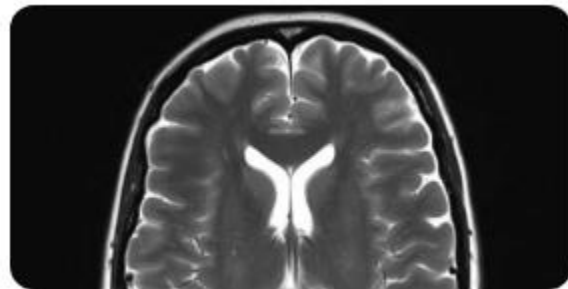
```
lookup.KeyValue  
f.constant(['em  
=tf.constant([0  
ce = tf.lookup.StaticV  
init,  
num_oov_buckets=5)  
  
lookup.StaticVocabular  
initializer,  
num_oov_buckets,  
lookup_key_dtype=None  
name=None,  
experimental_is_open
```

Code Review

Code Review

Brain tumors 256x256

A Refined Brain Tumor Image Dataset with Grayscale Normalization and Zoom



📁 glioma_tumor

2023-11-13 오후 11:56 파일 폴더

📁 meningioma_tumor

2023-11-13 오후 11:57 파일 폴더

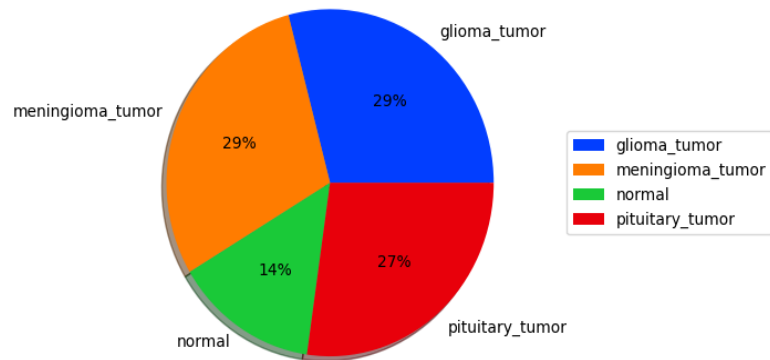
📁 normal

2023-11-13 오후 11:58 파일 폴더

📁 pituitary_tumor

2023-11-13 오후 11:59 파일 폴더

Data Preprocessing

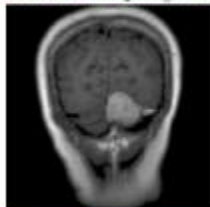


	Tumor_Name	glioma_tumor	meningioma_tumor	normal	pituitary_tumor	glioma_tumor_ratio	meningioma_tumor_ratio	normal_ratio	pituitary_tumor_ratio	SUM	Incorrect	Total	Incorrect Ratio
0	glioma_tumor	161	19	4	0	0.875	0.103261	0.021739	0	1	23	184	0.125
1	meningioma_tumor	9	161	0	9	0.050279	0.899441	0	0.050279	1	18	179	0.100559
2	normal	0	1	80	0	0	0.012346	0.987654	0	1	1	81	0.012346
3	pituitary_tumor	2	8	0	165	0.011429	0.045714	0	0.942857	1	10	175	0.057143

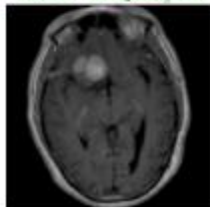
Actual: pituitary_tumor
Predicted: glioma_tumor



Actual: meningioma_tumor
Predicted: meningioma_tumor

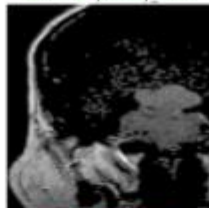


Actual: meningioma_tumor
Predicted: meningioma_tumor

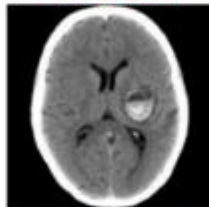


Actual: meningioma_tumor
Predicted: meningioma_tumor

Actual: pituitary_tumor
Predicted: pituitary_tumor



Actual: glioma_tumor
Predicted: normal

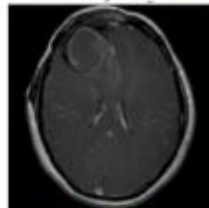


Actual: glioma_tumor
Predicted: glioma_tumor

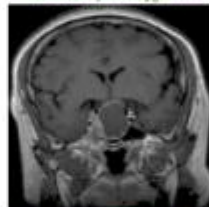


Actual: meningioma_tumor
Predicted: glioma_tumor

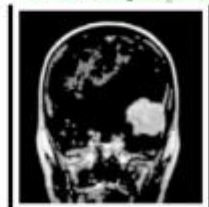
Actual: glioma_tumor
Predicted: glioma_tumor



Actual: pituitary_tumor
Predicted: pituitary_tumor

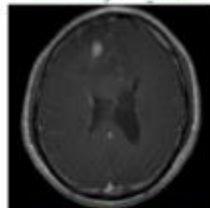


Actual: meningioma_tumor
Predicted: meningioma_tumor

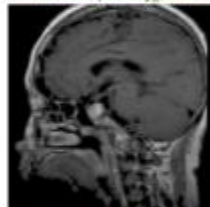


Actual: normal
Predicted: normal

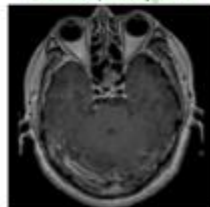
Actual: glioma_tumor
Predicted: glioma_tumor



Actual: pituitary_tumor
Predicted: pituitary_tumor



Actual: pituitary_tumor
Predicted: pituitary_tumor



Actual: glioma_tumor
Predicted: meningioma_tumor

Model

```
base_model =tf.keras.applications.resnet.ResNet50(include_top=False)

base_model.trainable = False


inputs = tf.keras.layers.Input(shape=(224, 224,3),name="input_layer")
x = base_model(inputs)

x =
tf.keras.layers.GlobalAveragePooling2D(name="global_average_pooling_layer"
)(x)

x=tf.keras.layers.Flatten()(x)
x=tf.keras.layers.Dense(512,
activation="relu")(x)
x=tf.keras.layers.Dropout(0.5)(x)

outputs=tf.keras.layers.Dense(4,
activation="softmax",name="output_layer")(x)

model= tf.keras.Model(inputs, outputs, name="model")
```



```
model.compile(loss=tf.keras.losses.SparseCategoricalCrossentropy(),
              optimizer=tf.keras.optimizers.Adam(lr=0.001),
              metrics=["accuracy"]
            )
```

```
learning_rate_reduction=tf.keras.callbacks.ReduceLROnPlateau(monitor=
"val_loss",patience=2,factor=0.5, min_lr=0.00001,verbose =1)
Early_Stopping=
tf.keras.callbacks.EarlyStopping(monitor="val_loss",patience=5,
restore_best_weights=True)
```

```
history1= model.fit(train_data, epochs=20, validation_data=
val_data,callbacks=[Early_Stopping,learning_rate_reduction], verbose=1)
```

Refactoring

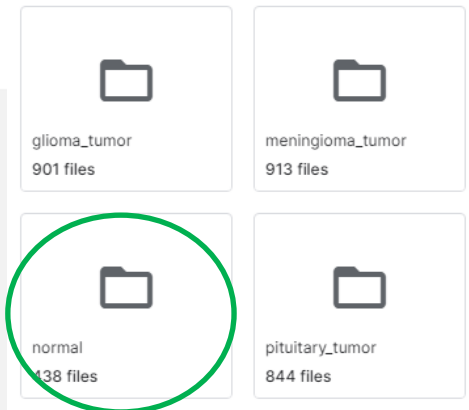
< 방향 설정 >

- 클래스 불균형 해결

- 모델 예측 결과 분석

- ResNet101, 152나 VGGNet, EfficientNet etc 활용

- 코드 재사용성 ↑



- 데이터 로딩 → 분할하는 처리 과정의 코드 재사용성 높이기
- model training 파트 코드 재사용성 높이기