```
[6]:
     import tensorflow as tf
     import numpy as np
      import matplotlib.pyplot as plt
      from sklearn.model_selection import train_test_split
      from tensorflow.keras import layers, models
      # Load and preprocess dataset
      full_ds = tf.keras.utils.image_dataset_from_directory(
         r'cat_dog',
                              # Root folder with subfolders as class names
         image_size=(228, 228),
         color_mode='grayscale'
                                                       # Any batch size works
      ).map(lambda x, y: (x / 255.0, y))
      # Convert dataset to numpy arrays
      x_full, y_full = [], []
      for images, labels in full ds
         x\_full.append(images.numpy())
         y_full.append(labels.numpy())
      x_full = np.concatenate(x_full, axis=0)
     y_full = np.concatenate(y_full, axis=0)
      # Train-test split
     x\_train, \ x\_test, \ y\_train, \ y\_test = train\_test\_split(x\_full, \ y\_full, \ test\_size=0.3, \ random\_state=42)
      # CNN model definition
     def build cnn model()
         inputs = tf.keras.Input(shape=(228, 228, 1))
         x = layers.Conv2D(16, 3, activation='relu')(inputs)
         x = layers.MaxPooling2D()(x)
         x = layers.Conv2D(32, 3, activation='relu')(x)
         x = layers.MaxPooling2D()(x)
         x = layers.Conv2D(64, 3, activation='relu')(x)
         x = layers.Flatten()(x)
         outputs = layers.Dense(10, activation='softmax')(x)\\
         model = models.Model(inputs=inputs, outputs=outputs)
         return model
      # Create the model
     model = build_cnn_model()
     model.summary()
      # Compile and train the model
     model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
     model.fit(x\_train,\ y\_train,\ epochs=5)
```

Found 132 files belonging to 2 classes.

[6]: Model: "functional_2"

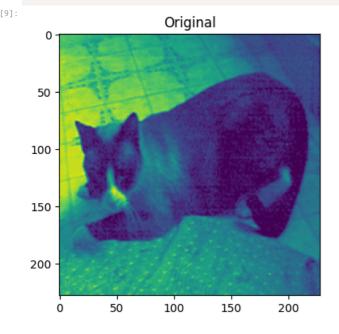
```
[6]:
                                                                     Τ
                                                                               Param #
     Layer (type)
                                         Output Shape
      input_layer_2 (InputLayer)
                                         ( No
                                                 228, 228, 1)
      conv2d_3 (Conv2D)
                                         (No
                                                 226, 226, 16)
                                                                     160
                                         (No
                                                                     max_pooling2d_2 (MaxPooling2D)
                                                 113, 113, 16)
                                                                                    0
                                                                     I (No
      conv2d_4 (Conv2D)
                                                                                 4.640 L
                                         (No
                                                                     0
      max_pooling2d_3 (MaxPooling2D)
                                                 55, 55, 32)
                                                                     conv2d_5 (Conv2D)
                                         (No
                                                                                18,496
                                                 53, 53, 64)
                                                                     flatten_1 (Flatten)
                                         (No
                                                 179776)
                                                                                    0
                                                                             1,797,770
                                         ( No
      dense_1 (Dense)
```

```
[6]: Total params: 1,821,066 (6.95 MB)
[6]: Trainable params: 1,821,066 (6.95 MB)
[6]: Non-trainable params: 0 (0.00 B)
[6]:
    Epoch 1/5
    3/3
                                7s 1s/step - accuracy: 0.2794 - loss: 1.7068
    Epoch 2/5
                                — 4s 1s/step - accuracy: 0.4744 - loss: 0.9120
     3/3
    Epoch 3/5
    3/3
                                4s 1s/step - accuracy: 0.6192 - loss: 0.6659
    Epoch 4/5
    3/3 -
                                - 4s 1s/step - accuracy: 0.5467 - loss: 0.7084
    Epoch 5/5
    3/3
                               — 4s 1s/step - accuracy: 0.7836 - loss: 0.5992
[6]:
```

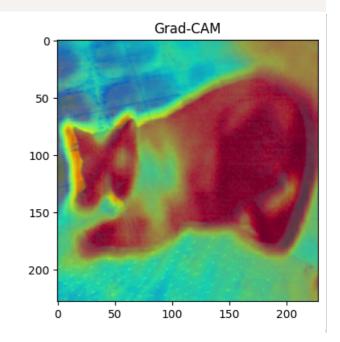
GradCAM

from tf_keras_vis.utils.scores import CategoricalScore from tf_keras_vis.utils.model_modifiers import ReplaceToLinear

```
from \ tensorflow.keras.preprocessing.image \ import \ load\_img,img\_to\_array
     # img_path = x_test[0]
     img = x_test[0]
     img_array = img_to_array(img) / 255.0
     img_input = np.expand_dims(img_array, axis=0)
     # 4. Apply Grad-CAM
     score = CategoricalScore([np.argmax(model.predict(img_input))])
     gm = Gradcam(model, model_modifier=ReplaceToLinear())
     cam = gm(score, img_input, penultimate_layer=-1)
     print(np.argmax(model.predict(img_input)),f"Original:{y_test[0]}")
[8]:
    1/1
                                 - Os 316ms/step
    1/1 -
                                 - Os 123ms/step
    1 Original:0
     import matplotlib.pyplot as plt
     f, ax = plt.subplots(1, 2, figsize=(12, 4))
     ax[0].imshow(img)
     ax[0].set_title("Original")
     ax[1].imshow(img\_array)
     ax[1].imshow(cam[0], cmap='jet', alpha=0.5)
     ax[1].set_title("Grad-CAM")
```



 $from \ tf_keras_vis.gradcam \ import \ Gradcam$

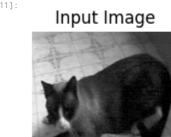


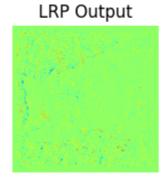
LRP AND SALIENCY MAP

plt.tight_layout()
plt.show()

From Test Data

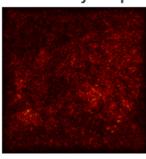
```
img = x_test[0]
img_tensor = tf.convert_to_tensor([img]) # Make it a batch
with tf.GradientTape() as tape:
    tape.watch(img_tensor)
    preds = model(img_tensor)
    class_idx = tf.argmax(preds[0])
    loss = preds[0, class_idx]
grads = tape.gradient(loss, img_tensor)[0].numpy()
relevance = grads * img # Gradient × Input relevance
saliency = np.max(np.abs(grads),axis=-1)
# Visualize input and relevance map
plt.subplot(1, 3, 1)
plt.imshow(img, cmap='gray')
plt.title("Input Image")
plt.axis('off')
plt.subplot(1, 3, 2)
plt.imshow(relevance,cmap='jet')
plt.title("LRP Output")
{\tt plt.axis('off')}
{\tt plt.show}(\,)
plt.subplot(1, 3, 3)
\verb"plt.imshow(saliency,cmap="hot")"
plt.title("Saliency Map")
plt.axis('off')
plt.show()
```







11]:



BY Custom Image

```
from tensorflow.keras.preprocessing.image import load_img,img_to_array
img=load_img('lion.jpg',target_size=(228, 228) # same image shape as in the model
                    , color\_mode='grayscale' #if (x,x,1) or if (x,x,3) no need
      img=img_to_array(img)/255.0
      img\_batch=np.expand\_dims(img,axis=0)
14]:
      img_tensor = tf.convert_to_tensor(img_batch) # Make it a batch
      with tf.GradientTape() as tape:
          {\tt tape.watch(img\_tensor)}
           preds = model(img_tensor)
          class_idx = tf.argmax(preds[0])
          loss = preds[0, class_idx]
      grads = tape.gradient(loss, img\_tensor)[0].numpy()
      relevance = grads * img # Gradient × Input relevance
      saliency = np.max(np.abs(grads), axis=-1)
      \ensuremath{\text{\#}} 
 Visualize input and relevance map
      plt.subplot(1,\ {\color{red}3},\ 1)
      plt.imshow(img, cmap='gray')
      plt.title("Input Image")
      plt.axis('off')
      plt.subplot(1,\ 3,\ 2)
      plt.imshow(relevance,cmap='jet')
      plt.title("LRP Output")
      plt.axis('off')
      plt.show()
      plt.subplot(1, 3, 3)
      plt.imshow(saliency,cmap='hot')
      plt.title("Saliency Map")
      plt.axis(\ensuremath{'off'})
      plt.show()
```

Input Image

14]:



LRP Output



