```
import tensorflow as tf
import os
from PIL import Image
import matplotlib.pyplot as plt
# Set dataset directory
dataset_dir = r"C:\Users\PMLS\Documents\deep learning\deep learning task 1\processed_images_grayscale"
# Load all image file paths
def load_image_paths(dataset_dir):
    image_paths = []
    for root, dirs, files in os.walk(dataset_dir):
        for file in files:
            if file.endswith(('jpg', 'jpeg', 'png')):
                image_paths.append(os.path.join(root, file))
    return image_paths
# Load and display an image
def load_and_display_image(image_path):
    img = Image.open(image_path)
    plt.imshow(img)
    plt.axis('off')
    plt.show()
# Example usage
image_paths = load_image_paths(dataset_dir)
# Display first image
if image_paths:
    print(f"Displaying image from: {image paths[0]}")
    load_and_display_image(image_paths[5666])
else:
    print("No images found in the dataset directory.")
```

Displaying image from: C:\Users\PMLS\Documents\deep learning\deep learning task 1\processed\_images\_grayscale\cat.0.jpg



```
import os
import cv2
import numpy as np
from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
# Load images and labels
def load_data():
    images = []
    labels = []
    for label in classes:
        class_dir = os.path.join(dataset_dir, label)
        for img_name in os.listdir(class_dir):
            img_path = os.path.join(class_dir, img_name)
           img = cv2.imread(img_path, cv2.IMREAD_GRAYSCALE)
            img = cv2.resize(img, img_size)
            images.append(img)
            labels.append(classes.index(label))
    return np.array(images), np.array(labels)
import os
import cv2
import numpy as np
from sklearn.model_selection import train_test_split
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense, Flatten
# Constants
dataset_dir = r"C:\Users\PMLS\Documents\deep learning\deep learning task 1\processed_images_grayscale"
img_size = (64, 64) # Resize images
# Load images and labels
def load_data():
    images = []
    labels = []
    for img_name in os.listdir(dataset_dir):
        img_path = os.path.join(dataset_dir, img_name)
        img = cv2.imread(img_path, cv2.IMREAD_GRAYSCALE)
        img = cv2.resize(img, img_size)
        images.append(img)
        # Infer label from filename
        if 'cat' in img name:
            labels.append(0) # Cat
        elif 'dog' in img_name:
           labels.append(1) # Dog
    return np.array(images), np.array(labels)
# Load data
X, y = load_data()
# Split the dataset
X = X.reshape(-1, img_size[0], img_size[1], 1) # Reshape for CNN
X = X.astype('float32') / 255.0 # Normalize pixel values
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Build the CNN model
model = models.Sequential()
# 1st Convolutional Layer
model.add(layers.Conv2D(32, (3, 3), activation='relu', input_shape=(300, 300, 1))) # Input shape matches image dimensions
model.add(layers.MaxPooling2D((2, 2)))
# 2nd Convolutional Layer
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
```

```
# 3rd Convolutional Layer
model.add(layers.Conv2D(128, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))

# Flatten the output from the Conv layers before passing it to fully connected layers
model.add(layers.Flatten())

# Fully connected (Dense) layer
model.add(layers.Dense(128, activation='relu'))

# Output layer (Assuming a multi-class classification problem)
model.add(layers.Dense(2, activation='softmax'))  # len(lb.classes_) gives the number of classes
```

C:\Users\PMLS\AppData\Local\Programs\Python\Python312\Lib\site-packages\keras\src\layers\convolutional\base\_conv.py:107: UserWarning: Dc
 super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs)

**←** 

model.summary()

→ Model: "sequential\_5"

Layer (type)	Output Shape	Param #
conv2d_15 (Conv2D)	(None, 298, 298, 32)	320
max_pooling2d_15 (MaxPooling2D)	(None, 149, 149, 32)	0
conv2d_16 (Conv2D)	(None, 147, 147, 64)	18,496
max_pooling2d_16 (MaxPooling2D)	(None, 73, 73, 64)	0
conv2d_17 (Conv2D)	(None, 71, 71, 128)	73,856
max_pooling2d_17 (MaxPooling2D)	(None, 35, 35, 128)	0
flatten_5 (Flatten)	(None, 156800)	0
dense_8 (Dense)	(None, 128)	20,070,528
dense_9 (Dense)	(None, 2)	258

Total params: 20,163,458 (76.92 MB)
Trainable params: 20,163,458 (76.92 MB)

```
→ Epoch 1/10
    625/625
                              - 62s 93ms/step - accuracy: 0.5873 - loss: 0.6624 - val_accuracy: 0.7252 - val_loss: 0.5499
    Epoch 2/10
    625/625 -
                             Epoch 3/10
    625/625 -
                             — 82s 92ms/step - accuracy: 0.8001 - loss: 0.4311 - val_accuracy: 0.8052 - val_loss: 0.4162
    Epoch 4/10
    625/625 -
                             — 57s 92ms/step - accuracy: 0.8337 - loss: 0.3721 - val_accuracy: 0.8104 - val_loss: 0.4038
    Epoch 5/10
    625/625 -
                             — 58s 93ms/step - accuracy: 0.8557 - loss: 0.3155 - val_accuracy: 0.8276 - val_loss: 0.3957
    Epoch 6/10
                             — 58s 92ms/step - accuracy: 0.8771 - loss: 0.2885 - val_accuracy: 0.8236 - val_loss: 0.3932
    625/625 -
    Epoch 7/10
    625/625 -
                             - 82s 93ms/step - accuracy: 0.9086 - loss: 0.2213 - val_accuracy: 0.8222 - val_loss: 0.4213
    Epoch 8/10
                             — 58s 93ms/step - accuracy: 0.9289 - loss: 0.1755 - val_accuracy: 0.8304 - val_loss: <b>0.4293
    625/625 -
    Epoch 9/10
    625/625 -
                             - 85s 97ms/step - accuracy: 0.9508 - loss: 0.1316 - val accuracy: 0.8292 - val loss: 0.5220
    Epoch 10/10
    625/625 -
                             — 60s 96ms/step - accuracy: 0.9626 - loss: 0.1016 - val_accuracy: 0.8192 - val_loss: 0.5782
moder.compile(optimizer='adam',
# Evaluate the model on the test set
test_loss, test_acc = model.evaluate(X_test, y_test, verbose=2)
```

157/157 - 8s - 53ms/step - accuracy: 0.8192 - loss: 0.5782 Test accuracy: 0.8191999793052673

Start coding or generate with AI.

print(f'Test accuracy: {test\_acc}')

Start coding or generate with AI.