```
!pip install -q kaggle
from google.colab import files
files.upload()
     Choose Files No file chosen
                                          Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
     enable.
!mkdir ~/.kaggle
!mv kaggle.json ~/.kaggle/
!chmod 600 ~/.kaggle/kaggle.json
!kaggle datasets download -d kaggleashwin/vehicle-type-recognition
Dataset URL: <a href="https://www.kaggle.com/datasets/kaggleashwin/vehicle-type-recognition">https://www.kaggle.com/datasets/kaggleashwin/vehicle-type-recognition</a>
     License(s): apache-2.0
     Downloading vehicle-type-recognition.zip to /content
      96% 153M/159M [00:01<00:00, 143MB/s]
     100% 159M/159M [00:01<00:00, 149MB/s]
!kaggle datasets download kaushalrimal619/lumpy-skin-disease-cow-images
Dataset URL: <a href="https://www.kaggle.com/datasets/kaushalrimal619/lumpy-skin-disease-cow-images">https://www.kaggle.com/datasets/kaushalrimal619/lumpy-skin-disease-cow-images</a>
     License(s): unknown
     Downloading lumpy-skin-disease-cow-images.zip to /content
     100% 4.27G/4.28G [00:58<00:00, 85.9MB/s]
     100% 4.28G/4.28G [00:58<00:00, 78.4MB/s]
!unzip lumpy-skin-disease-cow-images.zip
import tensorflow as tf
print("Num GPUs Available: ", len(tf.config.list_physical_devices('GPU')))
if tf.test.gpu device name():
    print(f"Using GPU: {tf.test.gpu_device_name()}")
else:
    print("No GPU found. Make sure you've enabled GPU in Colab.")
    Num GPUs Available: 0
     No GPU found. Make sure you've enabled GPU in Colab.
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormalization
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.callbacks import EarlyStopping
import os
import shutil
import matplotlib.pyplot as plt
infected_dir = '/content/infected/infected'
normal_dir = '/content/normal/normal'
def create_dataset(infected_dir, normal_dir):
    dataset_dir = '/content/dataset'
    train_dir = os.path.join(dataset_dir, 'train')
    os.makedirs(train dir, exist ok=True)
    infected_train_dir = os.path.join(train_dir, 'infected')
    normal_train_dir = os.path.join(train_dir, 'normal')
    os.makedirs(infected_train_dir, exist_ok=True)
    os.makedirs(normal_train_dir, exist_ok=True)
    for file in os.listdir(infected dir):
         shutil.copy(os.path.join(infected_dir, file), infected_train_dir)
```

```
for file in os.listdir(normal dir):
        shutil.copy(os.path.join(normal_dir, file), normal_train_dir)
   return dataset_dir
dataset_dir = create_dataset(infected_dir, normal_dir)
train_datagen = ImageDataGenerator(
   rescale=1.0 / 255,
   validation_split=0.2,
   rotation range=20,
   width_shift_range=0.2,
   height_shift_range=0.2,
   zoom_range=0.2,
   horizontal_flip=True
)
train_generator = train_datagen.flow_from_directory(
   os.path.join(dataset_dir, 'train'),
   target_size=(128, 128),
   batch_size=32,
   class_mode='binary',
   subset='training'
)
val_generator = train_datagen.flow_from_directory(
   os.path.join(dataset_dir, 'train'),
   target_size=(128, 128),
   batch_size=32,
   class_mode='binary',
   subset='validation'
Found 3200 images belonging to 2 classes.
     Found 800 images belonging to 2 classes.
model = Sequential([
   Conv2D(32, (3, 3), activation='relu', input shape=(128, 128, 3)),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2, 2)),
   Conv2D(64, (3, 3), activation='relu'),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2, 2)),
   Conv2D(128, (3, 3), activation='relu'),
   BatchNormalization(),
   MaxPooling2D(pool_size=(2, 2)),
   Flatten(),
   Dense(256, activation='relu'),
   Dropout(0.5),
   Dense(1, activation='sigmoid') # Output layer for binary classification
])
🚁 /usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:107: UserWarning: Do not pass an `input_shape`/`inpu
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
# Step 4: Compile the Model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
# Step 5: Train the Model
early_stop = EarlyStopping(monitor='val_loss', patience=5, restore_best_weights=True)
history = model.fit(
   train_generator,
   epochs=25,
   validation_data=val_generator,
   callbacks=[early_stop]
```

```
→ Epoch 1/25

     /usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:122: UserWarning: Your `PyDataset` class
       self._warn_if_super_not_called()
     100/100
                                 • 157s 1s/step - accuracy: 0.8976 - loss: 1.5282 - val_accuracy: 0.5000 - val_loss: 70.8111
     Epoch 2/25
     100/100
                                 - 132s 1s/step - accuracy: 0.9831 - loss: 0.1825 - val_accuracy: 0.5000 - val_loss: 77.4527
     Epoch 3/25
     100/100 -
                                  134s 1s/step - accuracy: 0.9876 - loss: 0.2018 - val_accuracy: 0.5038 - val_loss: 29.1804
     Epoch 4/25
                                  149s 1s/step - accuracy: 0.9862 - loss: 0.1574 - val_accuracy: 0.5487 - val_loss: 35.5958
     100/100
     Epoch 5/25
     100/100
                                  124s 1s/step - accuracy: 0.9926 - loss: 0.0601 - val_accuracy: 0.7275 - val_loss: 11.4234
     Epoch 6/25
                                  140s 1s/step - accuracy: 0.9950 - loss: 0.0513 - val_accuracy: 0.7400 - val_loss: 11.6773
     100/100
     Epoch 7/25
     100/100
                                  132s 1s/step - accuracy: 0.9908 - loss: 0.0958 - val_accuracy: 0.8425 - val_loss: 3.9482
     Enoch 8/25
     100/100
                                  124s 1s/step - accuracy: 0.9939 - loss: 0.0628 - val_accuracy: 0.7850 - val_loss: 8.9398
     Epoch 9/25
     100/100
                                  140s 1s/step - accuracy: 0.9911 - loss: 0.1514 - val_accuracy: 0.8562 - val_loss: 6.1548
     Epoch 10/25
     100/100
                                  142s 1s/step - accuracy: 0.9947 - loss: 0.0572 - val_accuracy: 0.6775 - val_loss: 20.4675
     Epoch 11/25
     100/100
                                  132s 1s/step - accuracy: 0.9920 - loss: 0.0654 - val_accuracy: 0.7937 - val_loss: 7.1895
     Epoch 12/25
     100/100
                                  133s 1s/step - accuracy: 0.9988 - loss: 0.0072 - val_accuracy: 0.8388 - val_loss: 4.7486
# Step 6: Save the Model
model.save('cow_health_classifier.h5')
wARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is consi
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['accuracy'], label='Training Accuracy')
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.legend()
plt.title('Accuracy')
plt.subplot(1, 2, 2)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.legend()
plt.title('Loss')
plt.show()
₹
                                                                                                       Loss
                                 Accuracy
                                                                          80
      1.0
                                                                                                                      Training Loss
                                                                                                                      Validation Loss
                                                                          70
      0.9
                                                                          60
                                                                          50
      0.8
                                                                          40
      0.7
                                                                          30
                                                                          20
      0.6
                                                                          10
                                              Training Accuracy
                                              Validation Accuracy
      0.5
                                                                           0
                                                          10
            0
                                                 8
    4
from google.colab import files
files.download('cow_health_classifier.h5')
```

```
from tensorflow.keras.models import load_model
model = load_model('cow_health_classifier.h5')
model.summary()
```

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile_metrics` will be empty until you t Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 126, 126, 32)	896
batch_normalization (BatchNormalization)	(None, 126, 126, 32)	128
max_pooling2d (MaxPooling2D)	(None, 63, 63, 32)	0
conv2d_1 (Conv2D)	(None, 61, 61, 64)	18,496
batch_normalization_1 (BatchNormalization)	(None, 61, 61, 64)	256
max_pooling2d_1 (MaxPooling2D)	(None, 30, 30, 64)	0
conv2d_2 (Conv2D)	(None, 28, 28, 128)	73,856
batch_normalization_2 (BatchNormalization)	(None, 28, 28, 128)	512
max_pooling2d_2 (MaxPooling2D)	(None, 14, 14, 128)	0
flatten (Flatten)	(None, 25088)	0
dense (Dense)	(None, 256)	6,422,784
dropout (Dropout)	(None, 256)	0
dense_1 (Dense)	(None, 1)	257

Total params: 6,517,187 (24.86 MB)

```
import matplotlib.pyplot as plt
from PIL import Image
def display_image(image_path):
   img = Image.open(image_path)
   plt.imshow(img)
   plt.axis('off')
   plt.show()
from tensorflow.keras.utils import load_img, img_to_array
import numpy as np
def preprocess_image(image_path):
   img = load_img(image_path, target_size=(128, 128)) # Resize the image to (128, 128)
   img_array = img_to_array(img) # Convert the image to a numpy array
   img_array = img_array / 255.0 # Normalize the pixel values to [0, 1]
   img_array = np.expand_dims(img_array, axis=0) # Add a batch dimension
   return img_array
image_path = '_/content/unhealthy_1.jpeg'
preprocessed_image = preprocess_image(image_path)
prediction = model.predict(preprocessed_image)
if prediction[0][0] > 0.5:
   print("Unhealthy Cow Detected (Infected)")
else:
   print("Healthy Cow Detected (Normal)")
display_image(image_path)
```

1/1 ______ 0s 18ms/step
Unhealthy Cow Detected (Infected)



Double-click (or enter) to edit

Double-click (or enter) to edit