# IMPORTANT: SOME KAGGLE DATA SOURCES ARE PRIVATE # RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES. import kagglehub kagglehub.login()



# kaggle

Create an API token from your Kaggle settings page and paste it below along with your Kaggle username.

Username:	
Token:	
	Login

**Thank You** 

#### Download Dataset

```
# IMPORTANT: RUN THIS CELL IN ORDER TO IMPORT YOUR KAGGLE DATA SOURCES,
# THEN FEEL FREE TO DELETE THIS CELL.
# NOTE: THIS NOTEBOOK ENVIRONMENT DIFFERS FROM KAGGLE'S PYTHON
# ENVIRONMENT SO THERE MAY BE MISSING LIBRARIES USED BY YOUR
# NOTEBOOK.
ismailnasri20_driver_drowsiness_dataset_ddd_path = kagglehub.dataset_download('ismailnasri20/driver-drowsiness-dataset-ddd')
print('Data source import complete.')
 Downloading from <a href="https://www.kaggle.com/api/v1/datasets/download/ismailnasri20/driver-drowsiness-dataset-ddd?dataset_version_number=1...">https://www.kaggle.com/api/v1/datasets/download/ismailnasri20/driver-drowsiness-dataset-ddd?dataset_version_number=1...</a>
      100% 22.5MB/s]Extracting files...
Imports
import os
import cv2
import numpy as np
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Sequential
```

ismailnasri20\_driver\_drowsiness\_dataset\_ddd\_path = kagglehub.dataset\_download('ismailnasri20/driver-drowsiness-dataset-ddd') print("Dataset path:", ismailnasri20\_driver\_drowsiness\_dataset\_ddd\_path) print("Contents:", os.listdir(ismailnasri20\_driver\_drowsiness\_dataset\_ddd\_path))

Dataset path: /root/.cache/kagglehub/datasets/ismailnasri20/driver-drowsiness-dataset-ddd/versions/1 Contents: ['Driver Drowsiness Dataset (DDD)']

from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout, BatchNormalization

from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint from tensorflow.keras.preprocessing.image import ImageDataGenerator

from tensorflow.keras.models import load\_model

from google.colab import files import matplotlib.pyplot as plt

import kagglehub

تقسيم البيانات #

from sklearn.model\_selection import train\_test\_split

```
def load_data(image_folder, img_size=(64, 64)):
    classes = {"Drowsy": 1, "Non Drowsy": 0}
   X, y = [], []
   base_folder = os.path.join(image_folder, "Driver Drowsiness Dataset (DDD)")
   print("Base folder:", base_folder)
   print("Base folder contents:", os.listdir(base_folder))
   for class_name in classes:
        class_dir = os.path.join(base_folder, class_name)
       try:
           images = os.listdir(class_dir)
           for img name in images:
               img_path = os.path.join(class_dir, img_name)
                img = cv2.imread(img_path)
               if img is not None:
                   img = cv2.resize(img, img_size)
                    X.append(img)
                   y.append(classes[class_name])
        except FileNotFoundError:
            print(f"Folder not found: {class_dir}")
           continue
   X = np.array(X)
   y = to_categorical(y)
   return X, y
```

```
تحميل البيانات #
dataset_path = ismailnasri20_driver_drowsiness_dataset_ddd_path
X, y = load_data(dataset_path)
```

Base folder: /root/.cache/kagglehub/datasets/ismailnasri20/driver-drowsiness-dataset-ddd/versions/1/Driver Drowsiness Dataset (DDD) Base folder contents: ['Drowsy', 'Non Drowsy']

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Data Augmentation
datagen = ImageDataGenerator(
   rotation_range=15,
   width_shift_range=0.1,
   height_shift_range=0.1,
   brightness_range=[0.8, 1.2],
   horizontal_flip=True
datagen.fit(X_train)
```

```
بناء النموذج #
model = Sequential()
model.add(Conv2D(32, (3,3), activation='relu', input_shape=(64, 64, 3)))
model.add(BatchNormalization())
```

```
model.add(MaxPooling2D(2,2))
model.add(Conv2D(64, (3,3), activation='relu'))
model.add(BatchNormalization())
model.add(MaxPooling2D(2,2))
model.add(Flatten())
model.add(Dense(128, activation='relu'))
model.add(BatchNormalization())
model.add(Dense(2, activation='softmax'))
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint, ReduceLROnPlateau
# 1. EarlyStopping
early_stop = EarlyStopping(
   monitor='val_loss',
   patience=5,
   min delta=0.001,
   restore_best_weights=True,
   verbose=1
# 2. ModelCheckpoint
checkpoint = ModelCheckpoint(
   'best_model.keras',
   monitor='val_accuracy',
   save_best_only=True,
   mode='max',
   verbose=1
# 3. ReduceLROnPlateau
reduce lr = ReduceLROnPlateau(
   monitor='val_loss',
   factor=0.2,
   patience=3,
   min_lr=0.00001,
   verbose=1
history = model.fit(
   datagen.flow(X_train, y_train, batch_size=32),
   epochs=30,
   validation_data=(X_test, y_test),
   callbacks=[early_stop, checkpoint, reduce_lr]
→ Epoch 1/30
    1045/1045 -
                             ─ 0s 54ms/step - accuracy: 0.9959 - loss: 0.0151
    Epoch 1: val_accuracy improved from -inf to 0.99904, saving model to best_model.keras
                             — 58s 55ms/step - accuracy: 0.9959 - loss: 0.0151 - val_accuracy: 0.9990 - val_loss: 0.0045 - learning_rate: 0.0010
    1045/1045 -
    Epoch 2/30
    1045/1045 -
                             ─ 0s 51ms/step - accuracy: 0.9943 - loss: 0.0180
    Epoch 2: val_accuracy did not improve from 0.99904
                             1045/1045 -
    Epoch 3/30
    1045/1045 -
                            -- 0s 50ms/step - accuracy: 0.9954 - loss: 0.0158
    Epoch 3: val_accuracy improved from 0.99904 to 0.99928, saving model to best_model.keras
    1045/1045 -
                             ── 54s 52ms/step - accuracy: 0.9954 - loss: 0.0158 - val_accuracy: 0.9993 - val_loss: 0.0023 - learning_rate: 0.0010
    Epoch 4/30
                              • 0s 50ms/step - accuracy: 0.9955 - loss: 0.0146
    1045/1045 -
    Epoch 4: val_accuracy did not improve from 0.99928
    1045/1045 -
                          Epoch 5/30
                            --- 0s 49ms/step - accuracy: 0.9948 - loss: 0.0167
    1045/1045 -
    Epoch 5: val accuracy did not improve from 0.99928
                            1045/1045 -
    Epoch 6/30
    1045/1045 -
                            --- 0s 50ms/step - accuracy: 0.9944 - loss: 0.0186
    Epoch 6: val_accuracy improved from 0.99928 to 0.99952, saving model to best_model.keras
    Epoch 6: ReduceLROnPlateau reducing learning rate to 0.00020000000949949026.
    1045/1045 -
                            —— 53s 51ms/step - accuracy: 0.9944 - loss: 0.0186 - val_accuracy: 0.9995 - val_loss: 0.0025 - learning_rate: 0.0010
    Epoch 7/30
    1044/1045 -
                            - 0s 49ms/step - accuracy: 0.9978 - loss: 0.0063
    Epoch 7: val_accuracy improved from 0.99952 to 0.99988, saving model to best_model.keras
                            — 53s 51ms/step - accuracy: 0.9978 - loss: 0.0063 - val_accuracy: 0.9999 - val_loss: 4.6419e-04 - learning_rate: 2.0000e-04
    1045/1045 ---
    Epoch 8/30
    1044/1045 -
                            −− 0s 50ms/step - accuracy: 0.9991 - loss: 0.0026
    Epoch 8: val_accuracy did not improve from 0.99988
                   1045/1045 ----
    Epoch 9/30
                            — 0s 49ms/step - accuracy: 0.9992 - loss: 0.0030
    Epoch 9: val_accuracy improved from 0.99988 to 1.00000, saving model to best_model.keras
                            —— 81s 50ms/step - accuracy: 0.9992 - loss: 0.0030 - val_accuracy: 1.0000 - val_loss: 5.4191e-04 - learning_rate: 2.0000e-04
    1045/1045 ----
    Epoch 10/30
    1045/1045 ----
                            --- 0s 51ms/step - accuracy: 0.9992 - loss: 0.0025
    Epoch 10: val_accuracy did not improve from 1.00000
    Epoch 10: ReduceLROnPlateau reducing learning rate to 4.0000001899898055e-05.
    1045/1045 -
                         Epoch 11/30
    1044/1045 -
                           --- 0s 50ms/step - accuracy: 0.9995 - loss: 0.0025
    Epoch 11: val_accuracy did not improve from 1.00000
    1045/1045 ---
                            —— 53s 51ms/step - accuracy: 0.9995 - loss: 0.0025 - val_accuracy: 0.9999 - val_loss: 4.4021e-04 - learning_rate: 4.0000e-05
    Epoch 12/30
    1045/1045 — Os 50ms/step - accuracy: 0.9995 - loss: 0.0017
    Epoch 12: val_accuracy did not improve from 1.00000
    1045/1045 ---
                             — 53s 50ms/step - accuracy: 0.9995 - loss: 0.0017 - val_accuracy: 0.9999 - val_loss: 3.9550e-04 - learning_rate: 4.0000e-05
    Epoch 12: early stopping
    Restoring model weights from the end of the best epoch: 7.
```

model.summary()

### → Model: "sequential\_2"

Layer (type)	Output Shape	Param #
conv2d_3 (Conv2D)	(None, 62, 62, 32)	896
batch_normalization (BatchNormalization)	(None, 62, 62, 32)	128
max_pooling2d_2 (MaxPooling2D)	(None, 31, 31, 32)	0
conv2d_4 (Conv2D)	(None, 29, 29, 64)	18,496
batch_normalization_1 (BatchNormalization)	(None, 29, 29, 64)	256
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 64)	0
flatten_1 (Flatten)	(None, 12544)	0
dense_2 (Dense)	(None, 128)	1,605,760
batch_normalization_2 (BatchNormalization)	(None, 128)	512
dense_3 (Dense)	(None, 2)	258

Total params: 4,878,024 (18.61 MB)
Trainable params: 1,625,858 (6.20 MB)
Non-trainable params: 448 (1.75 KB)

# Save
model.save('driverr\_drowsiness\_best\_model.keras')

loadedModel = load\_model('driverr\_drowsiness\_best\_model.keras')

```
from tensorflow.keras.models import load_model
from google.colab import files
loadedModel = load_model('driverr_drowsiness_best_model.keras')
def process_and_predict():
    uploaded = files.upload()
   for filename in uploaded.keys():
        img = cv2.imread(filename)
       if img is None:
           print("Error: Could not load the image. Please try another file.")
        img_resized = cv2.resize(img, (64, 64))
        img_input = np.expand_dims(img_resized, axis=0)
        prediction = loadedModel.predict(img_input)
        class_names = ["Non Drowsy", "Drowsy"]
        predicted_class = class_names[np.argmax(prediction)]
        img_rgb = cv2.cvtColor(img_resized, cv2.COLOR_BGR2RGB)
        plt.figure(figsize=(6, 4))
        plt.imshow(img_rgb)
        plt.title(f"Predicted class: {predicted_class}", fontsize=12, pad=10)
        plt.axis('off')
        plt.show()
```

process\_and\_predict()

## Predicted class: Drowsy



process\_and\_predict()

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.

Saving Drowsy-Driving (1).jpg to Drowsy-Driving (1) (3).jpg

1/1 ——— 0s 30ms/step

## Predicted class: Drowsy



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.

Saving 2-min-scaled.jpg to 2-min-scaled (1).jpg

1/1 ———— Os 30ms/step

#### Predicted class: Non Drowsy



process\_and\_predict()

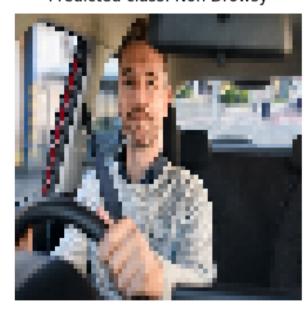
Choose Files No file chosen

Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

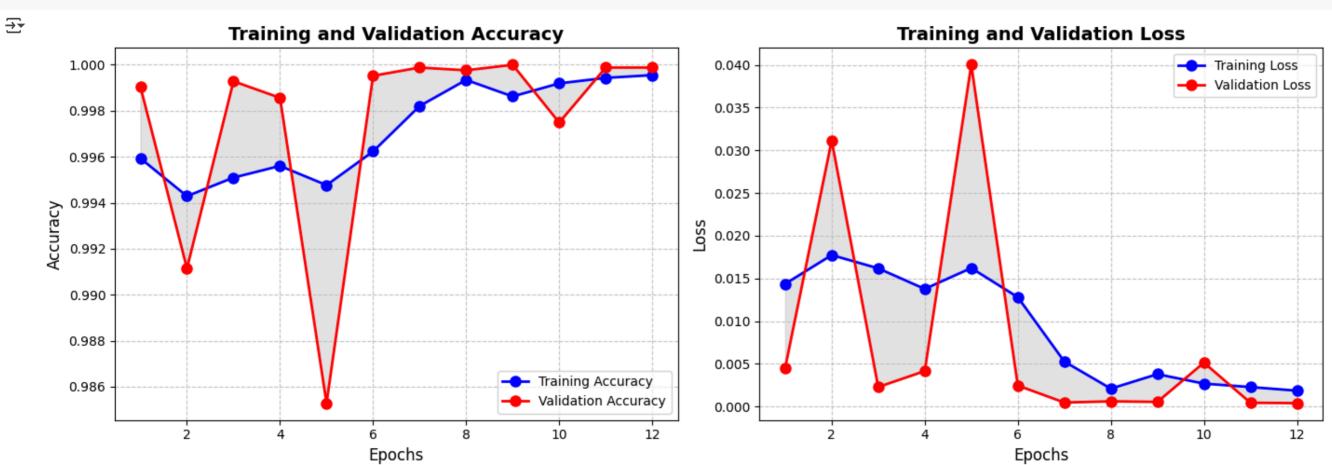
Saving man driving car=700x400.jpg to man-driving-car-700x400 (1).jpg

1/1 ———— 0s 29ms/step

Predicted class: Non Drowsy



```
def plot_training_history(history):
   acc = history.history['accuracy']
   val_acc = history.history['val_accuracy']
   loss = history.history['loss']
   val_loss = history.history['val_loss']
   epochs = range(1, len(acc) + 1)
   plt.figure(figsize=(14, 5))
   plt.subplot(1, 2, 1)
   plt.plot(epochs, acc, 'bo-', label='Training Accuracy', linewidth=2, markersize=8)
   plt.plot(epochs, val_acc, 'ro-', label='Validation Accuracy', linewidth=2, markersize=8)
   plt.title('Training and Validation Accuracy', fontsize=14, fontweight='bold')
   plt.xlabel('Epochs', fontsize=12)
   plt.ylabel('Accuracy', fontsize=12)
   plt.grid(True, linestyle='--', alpha=0.7)
   plt.legend(loc='lower right', fontsize=10)
   plt.fill_between(epochs, acc, val_acc, color='gray', alpha=0.2)
    plt.subplot(1, 2, 2)
    plt.plot(epochs, loss, 'bo-', label='Training Loss', linewidth=2, markersize=8)
    plt.plot(epochs, val_loss, 'ro-', label='Validation Loss', linewidth=2, markersize=8)
    plt.title('Training and Validation Loss', fontsize=14, fontweight='bold')
    plt.xlabel('Epochs', fontsize=12)
   plt.ylabel('Loss', fontsize=12)
   plt.grid(True, linestyle='--', alpha=0.7)
    plt.legend(loc='upper right', fontsize=10)
   plt.fill_between(epochs, loss, val_loss, color='gray', alpha=0.2)
   plt.tight_layout()
    plt.show()
plot_training_history(history)
```



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