## shoplifting-classification

## April 19, 2024

```
[13]: import pandas as pd
      import os
      import glob
      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 Conv2D, MaxPooling2D, Flatten, Dense
[14]: import os
      import pandas as pd
      def load_data(csv_file, video_folder):
          data = pd.read_csv(csv_file)
          video_names = data['video_name']
          labels = data['label']
          video_files = []
          for video_name in video_names:
              video_file = os.path.join(video_folder, video_name + '.mp4')
              video_files.append(video_file)
          return video_files, labels
[15]: import cv2
      import numpy as np
      def preprocess_data(video_paths, labels, target_height, target_width,_
       ⇔batch_size=32, augmentation=True):
          # Initialize lists to store frames and corresponding labels
          frames = []
          processed_labels = []
          for video_path, label in zip(video_paths, labels):
              # Read video
              cap = cv2.VideoCapture(video_path)
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while cap.isOpened():
        ret, frame = cap.read()
        if not ret:
            break
        # Resize frame
        frame = cv2.resize(frame, (target_width, target_height))
        # Data augmentation
        if augmentation:
            # Random horizontal flip
            if np.random.rand() < 0.5:</pre>
                frame = cv2.flip(frame, 1)
            # Random rotation (between -10 and 10 degrees)
            angle = np.random.randint(-10, 11)
            rows, cols, _ = frame.shape
            M = cv2.getRotationMatrix2D((cols / 2, rows / 2), angle, 1)
            frame = cv2.warpAffine(frame, M, (cols, rows))
            # Random brightness adjustment
            brightness = np.random.uniform(0.5, 1.5)
            frame = cv2.convertScaleAbs(frame, alpha=brightness)
        # Normalize frame
        frame = frame / 255.0 # Normalize pixel values to [0, 1]
        # Append frame and label
        frames.append(frame)
        processed_labels.append(label)
    cap.release()
# Convert lists to numpy arrays
frames = np.array(frames)
processed_labels = np.array(processed_labels)
return frames, processed_labels
model = models.Sequential()
model.add(layers.Conv2D(32, (3, 3), activation='relu', u
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[16]: def create_model(input_shape):
    model = models.Sequential()
    model.add(layers.Conv2D(32, (3, 3), activation='relu', u
    input_shape=input_shape))
    model.add(layers.MaxPooling2D((2, 2)))
    model.add(layers.Conv2D(64, (3, 3), activation='relu'))
    model.add(layers.MaxPooling2D((2, 2)))
    model.add(layers.Conv2D(128, (3, 3), activation='relu'))
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model.add(layers.MaxPooling2D((2, 2)))
         model.add(layers.Flatten())
         model.add(layers.Dense(128, activation='relu'))
         model.add(layers.Dense(1, activation='sigmoid'))
         return model
[17]: from sklearn.model_selection import train_test_split
     def split_data(frames, labels, test_size=0.2, batch_size=4):
          # split data into training and testing sets
         X_train, X_test, y_train, y_test = train_test_split(frames, labels, __
       # Calculate number of batches for training and testing
         train_batches = len(X_train) // batch_size
         test_batches = len(X_test) // batch_size
         # Ensure all data are divisible by batch size, discarding any remaining_
       ⇔samples
         X_train = X_train[:train_batches * batch_size]
         y_train = y_train[:train_batches * batch_size]
         X_test = X_test[:test_batches * batch_size]
         y_test = y_test[:test_batches * batch_size]
         return X_train, X_test, y_train, y_test
     csv_file = '/kaggle/input/csv-corrected/Shoplifting.csv'
     video_folder = '/kaggle/input/newshop/newshop'
     video_paths, labels = load_data(csv_file, video_folder)
     # Preprocess data
     target_height = 64
     target_width = 64
     frames, processed_labels = preprocess_data(video_paths, labels, target_height, u
       →target_width)
     print(frames)
     #Split data
     X_train, X_test, y_train, y_test = split_data(frames, processed_labels)
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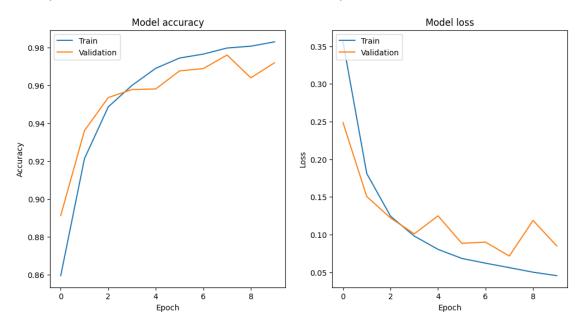
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[18]: import numpy as np
      import matplotlib.pyplot as plt
      import tensorflow as tf
      from sklearn.model_selection import train_test_split
      from tensorflow.keras import layers, models
      input_shape = (target_height, target_width, 3)
      model = create_model(input_shape)
      # compile the model
      model.compile(optimizer='adam',
                    loss='binary_crossentropy',
                    metrics=['accuracy'])
      # Training the model
      history = model.fit(X_train, y_train, epochs=10, batch_size=4,__
       ⇔validation_data=(X_test, y_test))
      # plot accuracy and loss curves
      plt.figure(figsize=(12, 6))
      # plot training & validation accuracy values
      plt.subplot(1, 2, 1)
      plt.plot(history.history['accuracy'])
      plt.plot(history.history['val_accuracy'])
      plt.title('Model accuracy')
      plt.ylabel('Accuracy')
      plt.xlabel('Epoch')
      plt.legend(['Train', 'Validation'], loc='upper left')
      # Plot training & validation loss values
      plt.subplot(1, 2, 2)
      plt.plot(history.history['loss'])
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plt.plot(history.history['val_loss'])
plt.title('Model loss')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Train', 'Validation'], loc='upper left')
plt.show()
/opt/conda/lib/python3.10/site-
packages/keras/src/layers/convolutional/base_conv.py:99: UserWarning: Do not
pass an `input_shape`/`input_dim` argument to a layer. When using Sequential
models, prefer using an `Input(shape)` object as the first layer in the model
instead.
  super().__init__(
Epoch 1/10
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accuracy: 0.7752 - loss: 0.6452
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accuracy: 0.8445 - loss: 0.4171
                              10898 graph_launch.cc:671] Fallback to op-by-op
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12288/12288
                        38s 3ms/step
- accuracy: 0.8445 - loss: 0.4170 - val_accuracy: 0.8913 - val_loss: 0.2487
Epoch 2/10
12288/12288
                       35s 3ms/step
- accuracy: 0.9114 - loss: 0.2043 - val_accuracy: 0.9361 - val_loss: 0.1505
Epoch 3/10
12288/12288
                        34s 3ms/step
- accuracy: 0.9466 - loss: 0.1274 - val_accuracy: 0.9535 - val_loss: 0.1221
Epoch 4/10
12288/12288
                        35s 3ms/step
- accuracy: 0.9581 - loss: 0.1024 - val_accuracy: 0.9578 - val_loss: 0.1009
Epoch 5/10
12288/12288
                       34s 3ms/step
- accuracy: 0.9685 - loss: 0.0807 - val_accuracy: 0.9581 - val_loss: 0.1248
Epoch 6/10
12288/12288
                       35s 3ms/step
- accuracy: 0.9746 - loss: 0.0670 - val_accuracy: 0.9676 - val_loss: 0.0884
Epoch 7/10
12288/12288
                       34s 3ms/step
- accuracy: 0.9783 - loss: 0.0588 - val_accuracy: 0.9688 - val_loss: 0.0899
Epoch 8/10
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                        35s 3ms/step
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[19]: # Evaluate the model on the test set
      from sklearn.metrics import classification_report, confusion_matrix
      import seaborn as sns
      y_pred = model.predict(X_test)
      y_pred_classes = np.argmax(y_pred, axis=1) # Assuming one-hot encoding for_
       ⇔predicted labels
      # classification report
      print(classification_report(y_test, y_pred_classes))
      # confusion matrix
      cm = confusion_matrix(y_test, y_pred_classes)
      plt.figure(figsize=(8, 6))
      sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', cbar=False)
      plt.title('Confusion Matrix')
      plt.xlabel('Predicted Labels')
      plt.ylabel('True Labels')
      plt.show()
```

384/384		1s 2ms	1s 2ms/step		
		precision	recall	f1-score	support
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accur	acy			0.84	12288
macro	avg	0.42	0.50	0.46	12288
weighted	avg	0.70	0.84	0.76	12288

/opt/conda/lib/python3.10/site-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

/opt/conda/lib/python3.10/site-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

/opt/conda/lib/python3.10/site-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero\_division` parameter to control this behavior.

\_warn\_prf(average, modifier, msg\_start, len(result))

