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Análise Exploratória dos Dados (EDA) dataset: https://mmlab.ie.cuhk.edu.hk/projects/CelebA.html

```
from google.colab import drive
import os
import matplotlib.pyplot as plt
from PIL import Image
drive.mount('/content/drive')
image_dir = '/content/drive/My Drive/fiap'
images = os.listdir(image_dir)

# Exibir algumas imagens
fig, axes = plt.subplots(1, 5, figsize=(20, 5))
for i, img_file in enumerate(images[:5]):
    img = Image.open(os.path.join(image_dir, img_file))
    axes[i].imshow(img)
    axes[i].axis('off')
plt.show()
```

⇒ Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force











## Pré-processamento de Dados

- Redimensionar e normalizar as imagens.
- Dividir os dados em conjuntos de treino e teste.

```
import numpy as np
from PIL import Image
from sklearn.model_selection import train_test_split
def load_images(image_dir, image_size=(128, 128)):
    images = []
    labels = []
    for img_file in os.listdir(image_dir):
        if ".jpg" in img_file:
             img = Image.open(os.path.join(image_dir, img_file)).resize(image_size)
             img_array = np.array(img)
             images.append(img_array)
             label = 1 if 'real' in img_file else 0
             labels.append(label)
    return np.array(images), np.array(labels)
image_dir = '/content/drive/My Drive/fiap'
images, labels = load_images(image_dir)
# Dividir em treino e teste
X_train, X_test, y_train, y_test = train_test_split(images, labels, test_size=0.25, random_state=42)
# Salvar dados processados
processed_dir = 'data/processed'
os.makedirs(processed_dir, exist_ok=True)
np.save(os.path.join(processed_dir, 'X_train.npy'), X_train)
np.save(os.path.join(processed_dir, 'X_test.npy'), X_test)
np.save(os.path.join(processed_dir, 'y_train.npy'), y_train)
np.save(os.path.join(processed_dir, 'y_test.npy'), y_test)
```

## Desenvolvimento do Modelo

· treinar e avaliar diferentes modelos.

```
import numpy as np
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.losses import BinaryCrossentropy
from sklearn.metrics import classification_report
# Carregar dados processados
X_train = np.load('data/processed/X_train.npy')
X_test = np.load('data/processed/X_test.npy')
y_train = np.load('data/processed/y_train.npy')
y_test = np.load('data/processed/y_test.npy')
# Normalizar os dados
X_train, X_test = X_train / 255.0, X_test / 255.0
# Definir o modelo
model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=(128, 128, 3)),
    MaxPooling2D((2, 2)),
    Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D((2, 2)),
    Flatten(),
    Dense(128, activation='relu'),
    Dense(1, activation='sigmoid')
])
# Compilar o modelo
model.compile(optimizer=Adam(), loss=BinaryCrossentropy(), metrics=['accuracy'])
# Treinar o modelo
model.fit(X_train, y_train, epochs=10, validation_data=(X_test, y_test))
y_pred = (model.predict(X_test) > 0.5).astype("int32")
print(classification_report(y_test, y_pred))
# Salvar o modelo
model.save('content/models/model.h5')
```

/usr/local/lib/python3.10/dist-packages/keras/src/layers/convolutional/base\_conv.py:107: UserWarning: Do not pass super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs) Epoch 1/10

```
- 24s 944ms/step - accuracy: 0.9236 - loss: 0.1063 - val_accuracy: 1.0000 - val_loss: 6.
24/24
Epoch 2/10
24/24
                         – 39s 863ms/step – accuracy: 1.0000 – loss: 1.1509e–30 – val accuracy: 1.0000 – val loss
Epoch 3/10
24/24 -
                          - 22s 890ms/step - accuracy: 1.0000 - loss: 1.3565e-30 - val_accuracy: 1.0000 - val_loss
Epoch 4/10
24/24
                          - 42s 933ms/step - accuracy: 1.0000 - loss: 4.8520e-31 - val_accuracy: 1.0000 - val_loss
Fnoch 5/10
                          - 40s 884ms/step - accuracy: 1.0000 - loss: 8.2421e-31 - val_accuracy: 1.0000 - val_loss
24/24
Epoch 6/10
24/24
                          - 40s 863ms/step - accuracy: 1.0000 - loss: 2.2237e-30 - val_accuracy: 1.0000 - val_loss
Epoch 7/10
                          - 41s 870ms/step - accuracy: 1.0000 - loss: 1.9788e-30 - val_accuracy: 1.0000 - val_loss
24/24
Epoch 8/10
                          - 41s 853ms/step - accuracy: 1.0000 - loss: 8.3267e-31 - val accuracy: 1.0000 - val loss
24/24 -
Epoch 9/10
24/24 -
                          - 22s 915ms/step - accuracy: 1.0000 - loss: 1.3094e-31 - val_accuracy: 1.0000 - val_loss
Epoch 10/10
                          - 20s 844ms/step - accuracy: 1.0000 - loss: 5.5239e-31 - val_accuracy: 1.0000 - val_loss
24/24
8/8
                        2s 193ms/step
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. Th
              precision
                           recall f1-score
                                              support
           0
                   1.00
                             1.00
                                       1.00
                                                  250
                                       1.00
                                                  250
   accuracy
                   1.00
                             1.00
                                       1.00
                                                  250
  macro avq
weighted avg
                   1.00
                             1.00
                                       1.00
                                                  250
```

## Avaliação do Modelo

• F1 Score

from sklearn.metrics import f1\_score
f1 = f1\_score(y\_test, y\_pred)
print(f'F1 Score: {f1}')

F1 Score: 0.0 /usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1760: UndefinedMetricWarning: F-score