

Validation de TP2&TP3 ~COMPUTER VISION~

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“FACE DETECTION APPLICATION”

- 1.Feature Engineering
- 2.Binary Classification using Scikit-Learn
- 3.Evaluating the Best Face Detector
- 4.Finding Faces in a New Image
- 5.Model Deployment with Python and Streamlit



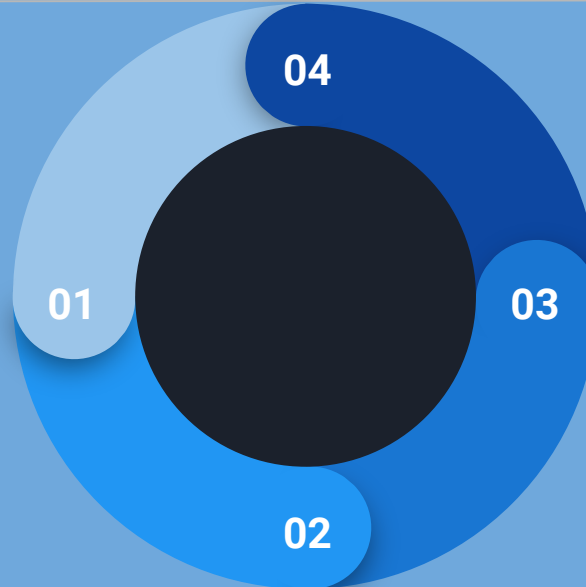
1. Feature Engineering

Dans un premier temps, nous réalisons feature engineering en extrayant l'histogramme d'orientation

Fonctions Gradients (HOG) à l'aide de la bibliothèque Scikit-Image.

sklearn

skimage



sklearn.feature_ext
raction.image

matplotlib.pyplot

2. Binary Classification using Scikit-Learn

Divisé les données en ensembles de formation et de test

```
# Split the data into training and test sets
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_s
tate=0)

# View the shape of the data
X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

Construisons différents classificateurs binaires avec le meilleur estimateur pour chacun :

```
from sklearn.ensemble import RandomForestClassifier
from sklearn import svm
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier

from sklearn.metrics import accuracy_score
from sklearn.model_selection import GridSearchCV
```


Create model

```
clf = model["estimator"]
```

Instantiate the grid search model

```
grid_search = GridSearchCV(estimator = clf, param_grid = model["params"],  
                           cv = 5)
```

Fit the model

```
grid_search.fit(X_train, y_train);
```

Make predictions on the test set compute accuracy metric

```
predicted = grid_search.predict(X_test)
```

```
acc = accuracy_score(predicted, y_test)
```

```
entries.append(acc)
```

```
print(grid_search.best_params_)
```

Get the best model with the highest accuracy

```
if acc > max_acc:
```

```
    max_acc = acc
```

```
    best_model = grid_search
```

```
import seaborn as sns
```

```
import pandas as pd
```

```
import matplotlib.pyplot as plt
```

```
plt.style.use('ggplot')
```

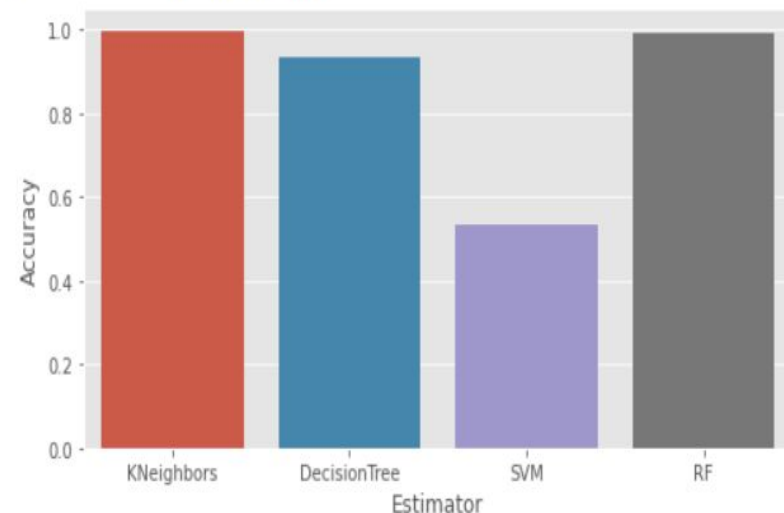
```
df = pd.DataFrame({"Estimator": column_names, "Accuracy": entries})
```

```
plt.figure(figsize=(8, 4))
```

```
sns.barplot(x='Estimator', y='Accuracy', data=df)
```

```
print(df)
```

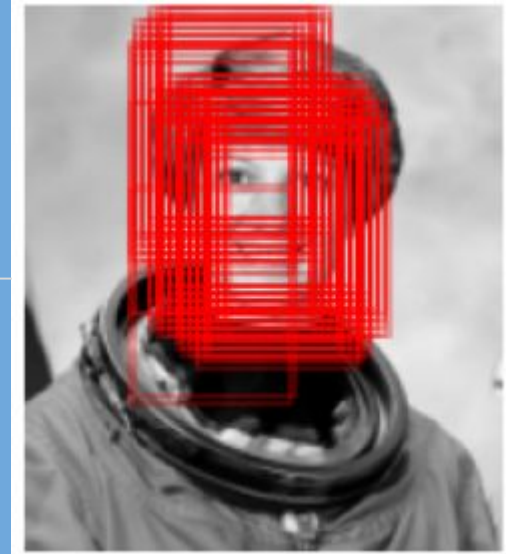
	Estimator	Accuracy
0	KNeighbors	0.997922
1	DecisionTree	0.936154
2	SVM	0.533245
3	RF	0.989611



Finding Faces in a New Image



`skimage.data`



`labels = best_model.predict(patches_hog)`



Face Detection App

Ichrak Ben Saad & Aya Fekih Romdhane

Face Detection

Upload File



Drag and drop file here

Limit 200MB per file • PNG, JPG, JPEG

Browse files



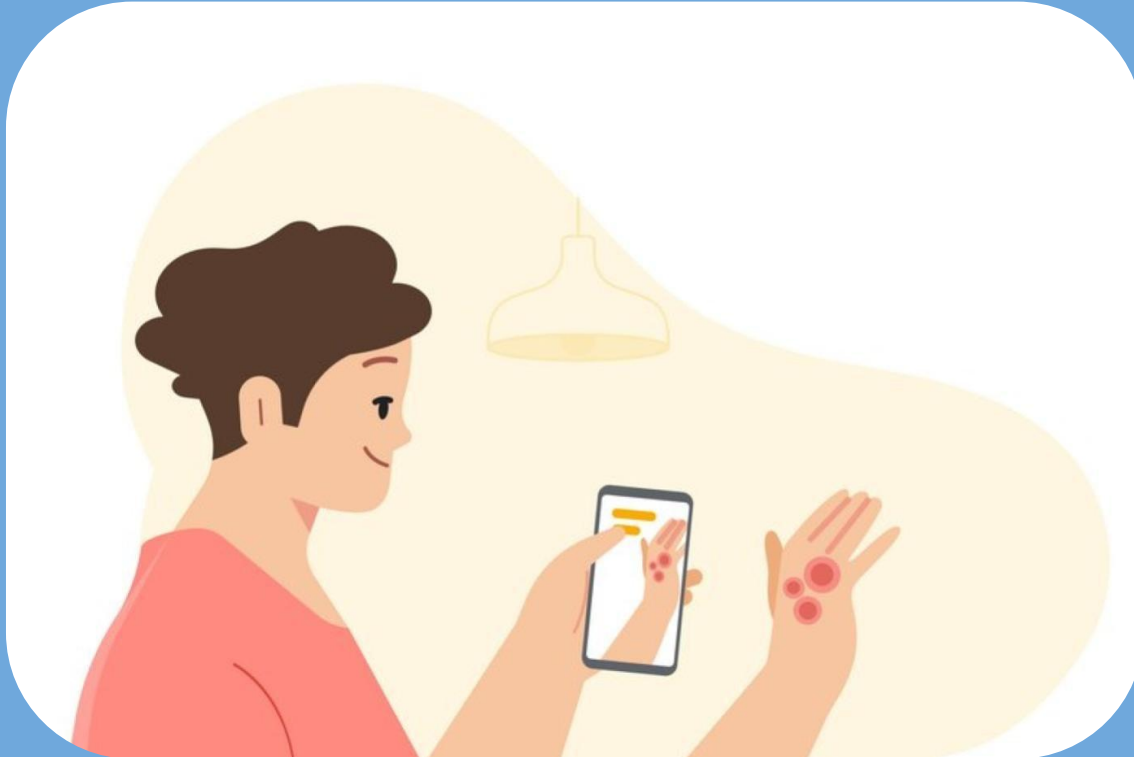
test3.PNG 28.8KB



Process



TP3 : DÉTECTION DU CANCER DE LA PEAU AVEC L'APPLICATION ANDROID



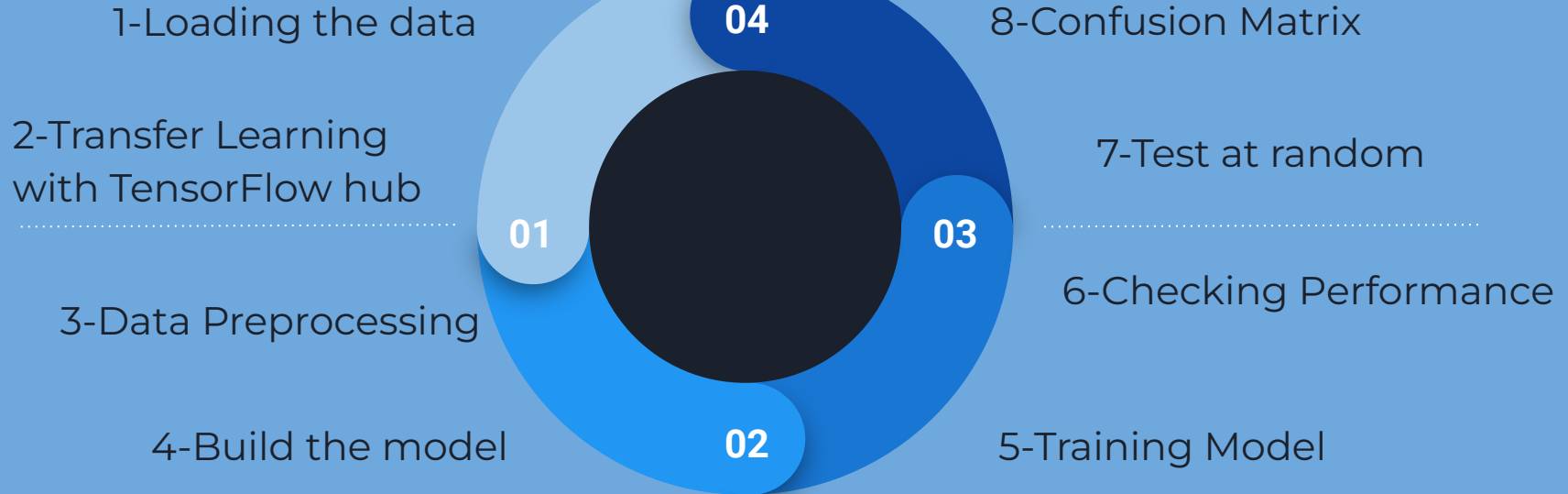


Le projet est principalement divisé en deux parties

- 01 **TensorFlow & Keras** sont utilisés pour construire et créer un modèle de machine learning.
- 02 **TFLite** est utilisé pour déployer le modèle sur une application Android.

Nos tâches pour la Training model

9-Export as saved model and convert to TFLite



Android App Part

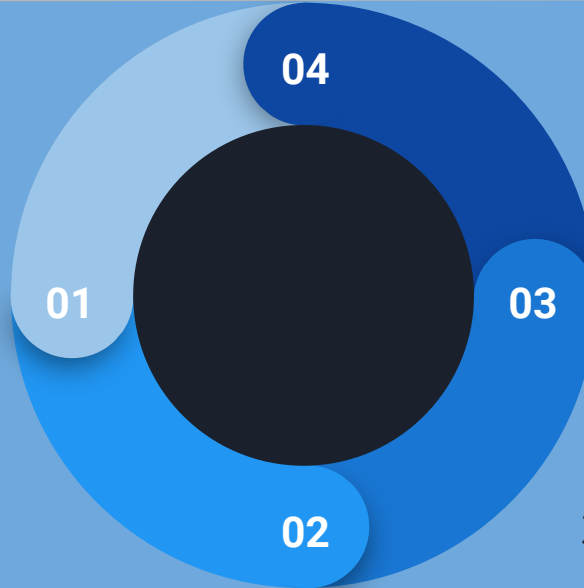
Nous utiliserons TFLite pour créer une application Android de bout en bout pour ce projet. Nous avons décidé de créer une application Android qui détecte le cancer de la peau.



Nos tâches pour l'application mobile

1-Load the model in our
Android project

2-Add TFLite
dependency to
app/build.gradle file



5-convert the
preprocessed bitmap into
ByteBuffer

4-Recognition is our
result data class

3-create a classifier

