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LAMPIRAN

1. Lampiran Listing Program

```
!python --version

!pip install --upgrade pip

!pip install mediapipe-model-maker

from google.colab import files

import os

import json

import tensorflow as tf

assert tf.__version__.startswith('2')

from mediapipe_model_maker import object_detector

#dataset format COCO

# !pip install roboflow

# from roboflow import Roboflow

# rf = Roboflow(api_key="YuVITDMltEsru3MPnh9U")

# project = rf.workspace("penyakit-bayam").project("penyakit-hama-
bayam")

# version = project.version(3)

# dataset = version.download("coco")

# !pip install roboflow

# from roboflow import Roboflow

# rf = Roboflow(api_key="YuVITDMltEsru3MPnh9U")

# project = rf.workspace("penyakit-bayam").project("penyakit-bayam")

# version = project.version(2)

# dataset = version.download("coco")

!pip install roboflow
```



```
from roboflow import Roboflow
```

```
Roboflow(api_key="YuVITDMltEsru3MPnh9U")
```

```

project = rf.workspace("penyakit-bayam").project("penyakit-bayam-new")
version = project.version(6)
dataset = version.download("coco")
train_dataset_path = "penyakit-bayam-new-6/train"
validation_dataset_path = "penyakit-bayam-new-6/valid"
test_dataset_path = "penyakit-bayam-new-6/test"

import os

import shutil

# Define the data directory
data_dir = train_dataset_path # Replace with the path to your data directory

# Define the target subdirectory
images_dir = os.path.join(data_dir, 'images')

# Create the target subdirectory if it doesn't exist
if not os.path.exists(images_dir):
    os.makedirs(images_dir)

# Move all jpg files to the images subdirectory
for filename in os.listdir(data_dir):
    if filename.endswith('.jpg'):
        source_path = os.path.join(data_dir, filename)
        target_path = os.path.join(images_dir, filename)
        shutil.move(source_path, target_path)

# Define the original and new filenames
original_filename = '_annotations.coco.json'
new_filename = 'labels.json'

# Construct full paths to the original and new files
original_filepath = os.path.join(data_dir, original_filename)
new_filepath = os.path.join(data_dir, new_filename)

# Check if the file exists
if os.path.exists(original_filepath):
    os.remove(original_filepath)

```



```

if os.path.exists(original_filepath):
    os.rename(original_filepath, new_filepath)
    print(f"File renamed from {original_filename} to {new_filename}")
else:
    print(f"File {original_filename} does not exist in the directory
    {data_dir}")

# Path to the labels.json file
labels_filepath = os.path.join(data_dir, 'labels.json')

# Check if the file exists
if not os.path.exists(labels_filepath):
    print(f"File {labels_filepath} does not exist.")
else:
    # Read the existing labels.json file
    with open(labels_filepath, 'r') as file:
        data = json.load(file)

    # Change the category with id 0 to "background"
    for category in data.get('categories', []):
        if category.get('id') == 0:
            category['name'] = 'background'
            break

    # Write the updated data back to labels.json
    with open(labels_filepath, 'w') as file:
        json.dump(data, file, indent=4)

    print("Label index 0 changed to 'background' successfully!")

# Define the data directory
_dir = validation_dataset_path # Replace with the path to your data
tory

fine the target subdirectory

```



```

images_dir = os.path.join(data_dir, 'images')

# Create the target subdirectory if it doesn't exist
if not os.path.exists(images_dir):
    os.makedirs(images_dir)

# Move all jpg files to the images subdirectory
for filename in os.listdir(data_dir):
    if filename.endswith('.jpg'):
        source_path = os.path.join(data_dir, filename)
        target_path = os.path.join(images_dir, filename)
        shutil.move(source_path, target_path)

# Define the original and new filenames
original_filename = '_annotations.coco.json'
new_filename = 'labels.json'

# Construct full paths to the original and new files
original_filepath = os.path.join(data_dir, original_filename)
new_filepath = os.path.join(data_dir, new_filename)

# Rename the file if it exists
if os.path.exists(original_filepath):
    os.rename(original_filepath, new_filepath)
    print(f"File renamed from {original_filename} to {new_filename}")

```



```

int(f"File {original_filename} does not exist in the directory
{data_dir}")

```

```

print("Files moved successfully!")

# Path to the labels.json file
labels_filepath = os.path.join(data_dir, 'labels.json')
# Check if the file exists
if not os.path.exists(labels_filepath):
    print(f"File {labels_filepath} does not exist.")
else:
    # Read the existing labels.json file
    with open(labels_filepath, 'r') as file:
        data = json.load(file)
    # Change the category with id 0 to "background"
    for category in data.get('categories', []):
        if category.get('id') == 0:
            category['name'] = 'background'
            break
    # Write the updated data back to labels.json
    with open(labels_filepath, 'w') as file:
        json.dump(data, file, indent=4)
    print("Label index 0 changed to 'background' successfully!")
# Define the data directory
data_dir = test_dataset_path # Replace with the path to your data directory

# Define the target subdirectory
images_dir = os.path.join(data_dir, 'images')

```

```

# Create the target subdirectory if it doesn't exist
if not os.path.exists(images_dir):
    os.makedirs(images_dir)

```



```

# Move all jpg files to the images subdirectory
for filename in os.listdir(data_dir):
    if filename.endswith('.jpg'):
        source_path = os.path.join(data_dir, filename)
        target_path = os.path.join(images_dir, filename)
        shutil.move(source_path, target_path)

# Define the original and new filenames
original_filename = '_annotations.coco.json'
new_filename = 'labels.json'

# Construct full paths to the original and new files
original_filepath = os.path.join(data_dir, original_filename)
new_filepath = os.path.join(data_dir, new_filename)

# Rename the file if it exists
if os.path.exists(original_filepath):
    os.rename(original_filepath, new_filepath)
    print(f"File renamed from {original_filename} to {new_filename}")
else:
    print(f"File {original_filename} does not exist in the directory {data_dir}")

print("Files moved successfully!")

# Path to the labels.json file
labels_filepath = os.path.join(data_dir, 'labels.json')

# Check if the file exists
if not os.path.exists(labels_filepath):
    print(f"File {labels_filepath} does not exist.")
else:
    Read the existing labels.json file
    with open(labels_filepath, 'r') as file:
        data = json.load(file)

```



```

# Change the category with id 0 to "background"
for category in data.get('categories', []):
    if category.get('id') == 0:
        category['name'] = 'background'
        break

# Write the updated data back to labels.json
with open(labels_filepath, 'w') as file:
    json.dump(data, file, indent=4)

print("Label index 0 changed to 'background' successfully!")

with open(os.path.join(train_dataset_path, "labels.json"), 'r') as f:
    labels_json = json.load(f)

for category_item in labels_json["categories"]:
    print(f"{category_item['id']}: {category_item['name']}")

train_data = object_detector.Dataset.from_coco_folder(train_dataset_path,
cache_dir="/tmp/od_data/train")

validation_data =
object_detector.Dataset.from_coco_folder(validation_dataset_path,
cache_dir="/tmp/od_data/validation")

test_data = object_detector.Dataset.from_coco_folder(test_dataset_path,
cache_dir="/tmp/od_data/test")

print("train_data size: ", train_data.size)

print("validation_data size: ", validation_data.size)

print("test_data size: ", test_data.size)

epoch_count = 130

spec = object_detector.SupportedModels.MOBILENET_MULTI_AVG

hparams = object_detector.HParams(export_dir='exported_model',
learning_rate=0.01,

```



```

    ochs=epoch_count)

ns = object_detector.ObjectDetectorOptions(
    pported_model=spec,

```

```

        hparams=hparams
    )
    model = object_detector.ObjectDetector.create(
        train_data=train_data,
        validation_data=validation_data,
        options=options)
    loss, coco_metrics = model.evaluate(validation_data, batch_size=4)
    print(f"Validation loss: {loss}")
    print(f"Validation coco metrics: {coco_metrics}")
    model.export_model()
    # !ls exported_model
    # files.download('exported_model/model.tflite')
    from mediapipe_model_maker import quantization
    model.restore_float_ckpt()
    model.export_model(model_name="model_fp16.tflite",
        quantization_config=quantization_config)
    !ls -lh exported_model
    files.download('exported_model/model_fp16.tflite')

```

