

Data Arguments

January 5, 2024

```
[ ]: pip install --upgrade keras
[6]: from tensorflow import keras
[ ]: pip show tensorflow
[ ]: pip show keras
[ ]: pip install tensorflow
[27]: #from tensorflow.keras.preprocessing.image import ImageDataGenerator,
       ↪array_to_img, img_to_array, load_img
#from keras.preprocessing.image import ImageDataGenerator , array_to_img ,,
       ↪img_to_array , load_img

from tensorflow.keras.utils import image_dataset_from_directory
from keras.utils import array_to_img ,img_to_array, load_img

datagen = ImageDataGenerator(
    rotation_range=40,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    fill_mode='nearest')

[4]: import tensorflow.keras as keras
[28]: img = load_img(r"C:\Users\srava\Downloads\mahindra-thar-2020-black.jpg")
[29]: img
[29]:
```



```
[30]: x = img_to_array(img) # this is a Numpy array with shape (3, 150, 150)
x = x.reshape((1,) + x.shape) # this is a Numpy array with shape (1, 3, 150, 150)

# the .flow() command below generates batches of randomly transformed images
# and saves the results to the `preview/` directory
i = 0
for batch in datagen.flow(x, batch_size=1,
                           save_to_dir=r"C:\Users\srava\OneDrive\Desktop\Datasets\Arguments",
                           save_prefix='cat', save_format='jpeg'):
    i += 1
    if i > 30:
        break # otherwise the generator would loop indefinitely
```

```
[24]: from tensorflow.keras.utils import image_dataset_from_directory
from keras.utils import array_to_img ,img_to_array, load_img

datagen = ImageDataGenerator(
    rotation_range=40,
    width_shift_range=0.2,
    height_shift_range=0.2,
    shear_range=0.2,
    zoom_range=0.2,
    horizontal_flip=True,
    fill_mode='reflect')
img = load_img(r"C:\Users\srava\Downloads\mahindra-thar-2020-black.jpg")
```

```
img
```

[24] :



```
[26]: x = img_to_array(img) # this is a Numpy array with shape (3, 150, 150)
x = x.reshape((1,) + x.shape) # this is a Numpy array with shape (1, 3, 150, 150)

# the .flow() command below generates batches of randomly transformed images
# and saves the results to the `preview/` directory
i = 0
for batch in datagen.flow(x, batch_size=1,
                           save_to_dir=r"C:\Users\sarva\OneDrive\Desktop\data\"
                           ↪Arguments Image creation", save_prefix='cat', save_format='jpeg'):
    i += 1
    if i > 30:
        break # otherwise the generator would loop indefinitely
```

```
[12]: keras.applications.ResNet50(
        include_top=True,
        weights="imagenet",
        input_tensor=None,
        input_shape=None,
        pooling=None,
        classes=1000,
        classifier_activation="softmax",
    )
```

Downloading data from <https://storage.googleapis.com/tensorflow/keras->

```
applications/resnet/resnet50_weights_tf_dim_ordering_tf_kernels.h5  
102967424/102967424 [=====] - 8s 0us/step
```

```
[12]: <keras.engine.functional.Functional at 0x21885b9a8f0>
```

1 Basic Predictions of the Image

2 RESNET50

```
[7]: from keras.utils import array_to_img ,img_to_array, load_img  
img = load_img(r"C:\Users\srava\Downloads\mahindra-thar-2020-black.jpg")  
img
```

```
[7]:
```



```
[32]: import numpy as np  
from tensorflow.keras.applications.resnet50 import ResNet50, preprocess_input,decode_predictions  
from tensorflow.keras.preprocessing import image  
  
# Load the ResNet-50 model pre-trained on ImageNet data  
model = ResNet50(weights='imagenet')  
  
# Load and preprocess the input image  
img_path = r"C:\Users\srava\Downloads\mahindra-thar-2020-black.jpg" # replace  
# with the path to your image file  
img = image.load_img(img_path, target_size=(224, 224))  
img_array = image.img_to_array(img)
```

```



```

```

1/1 [=====] - 2s 2s/step
Downloading data from https://storage.googleapis.com/download.tensorflow.org/data/imagenet_class_index.json
35363/35363 [=====] - 0s 0us/step
Predictions:
1: jeep (1.00)
2: pickup (0.00)
3: grille (0.00)

```

Top Prediction Class Index: 609

3 ResNet50V2

```

[36]: import numpy as np
from tensorflow.keras.applications.resnet_v2 import ResNet50V2, ↴
    preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the ResNet50V2 model pre-trained on ImageNet data
model = ResNet50V2(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\mahindra-thar-2020-black.jpg" # replace ↴
    with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions

```

```

predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=3)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50v2_weights_tf_dim_ordering_tf_kernels.h5
102869336/102869336 [=====] - 8s 0us/step
1/1 [=====] - 1s 1s/step
Predictions:
1: jeep (1.00)
2: pickup (0.00)
3: grille (0.00)

Top Prediction Class Index: 609

4 VGG16

```

[37]: import numpy as np
from tensorflow.keras.applications.vgg16 import VGG16, preprocess_input,
decode_predictions
from tensorflow.keras.preprocessing import image

# Load the VGG16 model pre-trained on ImageNet data
model = VGG16(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\sarav\Downloads\mahindra-thar-2020-black.jpg" # replace
with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions
predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=3)[0]

```

```
print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

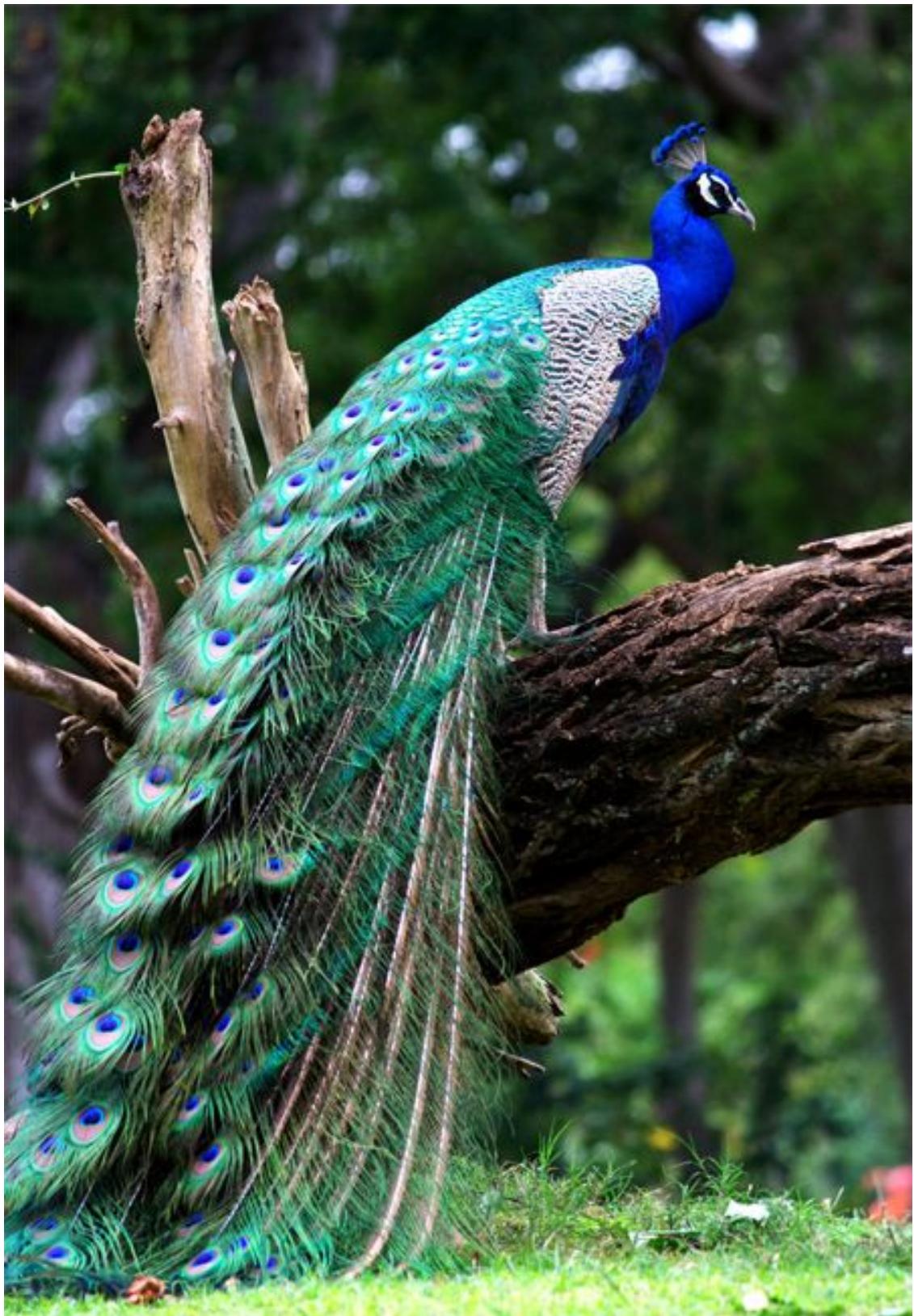
# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5
553467096/553467096 [=====] - 42s 0us/step
1/1 [=====] - 1s 555ms/step
Predictions:
1: jeep (0.99)
2: plow (0.00)
3: tractor (0.00)

Top Prediction Class Index: 609

```
[8]: from keras.utils import array_to_img ,img_to_array, load_img
img = load_img(r"C:\Users\srava\Downloads\peacock.jpg")
img
```

[8]:



```
[1]: import numpy as np
from tensorflow.keras.applications.vgg16 import VGG16, preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the VGG16 model pre-trained on ImageNet data
model = VGG16(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\peacock.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions
predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=3)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")
```

1/1 [=====] - 1s 699ms/step

Predictions:

1: peacock (1.00)
 2: indigo_bunting (0.00)
 3: hummingbird (0.00)

Top Prediction Class Index: 84

5 VGG19

```
[9]: from keras.utils import array_to_img ,img_to_array, load_img
img = load_img(r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg")
img
```

[9]:



```
[2]: import numpy as np
from tensorflow.keras.applications.vgg19 import VGG19, preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the VGG19 model pre-trained on ImageNet data
model = VGG19(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions
predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=3)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")
```

```
# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg19/vgg19_weights_tf_dim_ordering_tf_kernels.h5
574710816/574710816 [=====] - 246s 0us/step
1/1 [=====] - 1s 583ms/step
Predictions:
1: golden_retriever (0.78)
2: Saluki (0.13)
3: Afghan_hound (0.03)

Top Prediction Class Index: 207

```
[4]: import numpy as np
from tensorflow.keras.applications.vgg19 import VGG19, preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the VGG19 model pre-trained on ImageNet data
model = VGG19(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions
predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=8)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")
```

1/1 [=====] - 1s 599ms/step
Predictions:
1: golden_retriever (0.78)

```
2: Saluki (0.13)
3: Afghan_hound (0.03)
4: standard_poodle (0.02)
5: Labrador_retriever (0.01)
6: kuvasz (0.01)
7: redbone (0.00)
8: Irish_setter (0.00)
```

Top Prediction Class Index: 207

```
[5]: import numpy as np
from tensorflow.keras.applications.resnet_v2 import ResNet50V2,_
    preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the ResNet50V2 model pre-trained on ImageNet data
model = ResNet50V2(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the_
    ↪path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Make predictions
predictions = model.predict(img_array)

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=10)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")
```

```
1/1 [=====] - 2s 2s/step
Predictions:
1: golden_retriever (1.00)
2: Labrador_retriever (0.00)
3: kuvasz (0.00)
4: cocker_spaniel (0.00)
5: redbone (0.00)
6: Great_Pyrenees (0.00)
```

```
7: Tibetan_mastiff (0.00)
8: English_setter (0.00)
9: Leonberg (0.00)
10: Newfoundland (0.00)
```

Top Prediction Class Index: 207

6 Predictions for dog image using Keras Api Applications.....

```
[49]: from keras.utils import array_to_img ,img_to_array, load_img
img = load_img(r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg")
img
```

[49] :



```
[21]: import numpy as np
from tensorflow.keras.applications.resnet50 import ResNet50, preprocess_input,decode_predictions
from tensorflow.keras.preprocessing import image

# Load the ResNet-50 model pre-trained on ImageNet data
model = ResNet50(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
```

```

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

1/1 [=====] - 1s 732ms/step
 Predictions:
 1: golden_retriever (0.99)
 2: Labrador_retriever (0.00)
 3: kuvasz (0.00)
 4: Afghan_hound (0.00)
 5: standard_poodle (0.00)

Top Prediction Class Index: 207

```
Inference Time: 765.26 ms
Size (MB): 97.80 MB
Parameters: 25636712
Depth: 177
```

```
[22]: import numpy as np
from tensorflow.keras.applications.resnet_v2 import ResNet50V2, u
    ↪preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the ResNet50V2 model pre-trained on ImageNet data
model = ResNet50V2(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the u
    ↪path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"\nInference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"\nSize (MB): {model_size_MB:.2f} MB")
```

```
# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")
```

1/1 [=====] - 1s 1s/step

Predictions:

1: golden_retriever (1.00)
 2: Labrador_retriever (0.00)
 3: kuvasz (0.00)
 4: cocker_spaniel (0.00)
 5: redbone (0.00)

Top Prediction Class Index: 207

Inference Time: 1275.22 ms

Size (MB): 97.71 MB

Parameters: 25613800

Depth: 192

```
[25]: import numpy as np
from tensorflow.keras.applications.vgg19 import VGG19, preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image
import time

# Load the VGG19 model pre-trained on ImageNet data
model = VGG19(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
```

```

decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

1/1 [=====] - 0s 209ms/step

Predictions:

- 1: golden_retriever (0.78)
- 2: Saluki (0.13)
- 3: Afghan_hound (0.03)
- 4: standard_poodle (0.02)
- 5: Labrador_retriever (0.01)

Top Prediction Class Index: 207

Inference Time: 233.69 ms

Size (MB): 548.05 MB

Parameters: 143667240

Depth: 26

```
[26]: import numpy as np
from tensorflow.keras.applications.vgg16 import VGG16, preprocess_input,
decode_predictions
from tensorflow.keras.preprocessing import image
```

```

# Load the VGG16 model pre-trained on ImageNet data
model = VGG16(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the
# path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

1/1 [=====] - 0s 181ms/step

```
Predictions:  
1: golden_retriever (0.89)  
2: Afghan_hound (0.05)  
3: Saluki (0.02)  
4: bloodhound (0.01)  
5: Labrador_retriever (0.01)
```

```
Top Prediction Class Index: 207  
Inference Time: 214.87 ms  
Size (MB): 527.79 MB  
Parameters: 138357544  
Depth: 23
```

```
[28]: import numpy as np  
from tensorflow.keras.applications import Xception  
from tensorflow.keras.preprocessing import image  
from tensorflow.keras.applications.xception import preprocess_input,  
    decode_predictions  
  
# Load the Xception model pre-trained on ImageNet data  
model = Xception(weights='imagenet')  
  
# Load and preprocess the input image  
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the  
    ↪path to your image file  
img = image.load_img(img_path, target_size=(299, 299)) # Xception requires  
    ↪input shape (299, 299)  
img_array = image.img_to_array(img)  
img_array = np.expand_dims(img_array, axis=0)  
img_array = preprocess_input(img_array)  
  
# Measure inference time  
start_time = time.time()  
predictions = model.predict(img_array)  
end_time = time.time()  
  
# Decode and print the top-3 predicted classes  
decoded_predictions = decode_predictions(predictions, top=5)[0]  
  
print("Predictions:")  
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):  
    print(f"{i + 1}: {label} ({score:.2f})")  
  
# Optionally, you can obtain the class index for the top prediction  
top_class_index = np.argmax(predictions[0])  
print(f"\nTop Prediction Class Index: {top_class_index}")
```

```

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/xception/xception_weights_tf_dim_ordering_tf_kernels.h5
91884032/91884032 [=====] - 26s 0us/step
1/1 [=====] - 2s 2s/step
Predictions:
1: golden_retriever (0.97)
2: Labrador_retriever (0.01)
3: cocker_spaniel (0.00)
4: tennis_ball (0.00)
5: kuvasz (0.00)

Top Prediction Class Index: 207
Inference Time: 1623.19 ms
Size (MB): 87.40 MB
Parameters: 22910480
Depth: 134

```
[30]: import numpy as np
from tensorflow.keras.applications import InceptionV3
from tensorflow.keras.preprocessing import image
from tensorflow.keras.applications.inception_v3 import preprocess_input,decode_predictions

# Load the InceptionV3 model pre-trained on ImageNet data
model = InceptionV3(weights='imagenet')

# Load and preprocess the input image
```

```



```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/inception_v3/inception_v3_weights_tf_dim_ordering_tf_kernels.h5
96112376/96112376 [=====] - 20s 0us/step
1/1 [=====] - 2s 2s/step

```
Predictions:  
1: golden_retriever (0.95)  
2: Labrador_retriever (0.01)  
3: tennis_ball (0.00)  
4: doormat (0.00)  
5: cocker_spaniel (0.00)
```

```
Top Prediction Class Index: 207  
Inference Time: 2438.82 ms  
Size (MB): 90.99 MB  
Parameters: 23851784  
Depth: 313
```

```
[33]: import numpy as np  
from tensorflow.keras.applications import MobileNetV2  
from tensorflow.keras.preprocessing import image  
from tensorflow.keras.applications.mobilenet_v2 import preprocess_input,  
    decode_predictions  
  
# Load the MobileNetV2 model pre-trained on ImageNet data  
model = MobileNetV2(weights='imagenet')  
  
# Load and preprocess the input image  
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the  
    ↪path to your image file  
img = image.load_img(img_path, target_size=(224, 224))  
img_array = image.img_to_array(img)  
img_array = np.expand_dims(img_array, axis=0)  
img_array = preprocess_input(img_array)  
  
# Measure inference time  
start_time = time.time()  
predictions = model.predict(img_array)  
end_time = time.time()  
  
# Decode and print the top-3 predicted classes  
decoded_predictions = decode_predictions(predictions, top=5)[0]  
  
print("Predictions:")  
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):  
    print(f"{i + 1}: {label} ({score:.2f})")  
  
# Optionally, you can obtain the class index for the top prediction  
top_class_index = np.argmax(predictions[0])  
print(f"\nTop Prediction Class Index: {top_class_index}")  
  
# Calculate and print the inference time per step
```

```

inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

1/1 [=====] - 1s 1s/step

Predictions:

- 1: golden_retriever (0.97)
- 2: Labrador_retriever (0.01)
- 3: Chesapeake_Bay_retriever (0.00)
- 4: cocker_spaniel (0.00)
- 5: flat-coated_retriever (0.00)

Top Prediction Class Index: 207

Inference Time: 1306.48 ms

Size (MB): 13.50 MB

Parameters: 3538984

Depth: 156

```
[38]: import numpy as np
from tensorflow.keras.applications import DenseNet121
from tensorflow.keras.applications.densenet import preprocess_input, ▾
    decode_predictions
from tensorflow.keras.preprocessing import image

# Load the DenseNet121 model pre-trained on ImageNet data
model = DenseNet121(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the ▾
    path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
```

```

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/densenet/densenet121_weights_tf_dim_ordering_tf_kernels.h5
33188688/33188688 [=====] - 11s 0us/step
1/1 [=====] - 3s 3s/step
Predictions:
1: golden_retriever (0.87)
2: Labrador_retriever (0.11)
3: cocker_spaniel (0.00)
4: kuvasz (0.00)
5: Great_Pyrenees (0.00)

```
Top Prediction Class Index: 207
Inference Time: 3114.43 ms
Size (MB): 30.76 MB
Parameters: 8062504
Depth: 429
```

```
[42]: import numpy as np
from tensorflow.keras.applications import NASNetMobile
from tensorflow.keras.applications.nasnet import preprocess_input, decode_predictions
from tensorflow.keras.preprocessing import image

# Load the NASNetMobile model pre-trained on ImageNet data
model = NASNetMobile(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(224, 224))
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
```

```

model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/nasnet/NASNet-mobile.h5
24227760/24227760 [=====] - 9s 0us/step
1/1 [=====] - 6s 6s/step
Predictions:
1: golden_retriever (0.91)
2: Labrador_retriever (0.02)
3: redbone (0.01)
4: tennis_ball (0.00)
5: Irish_setter (0.00)

Top Prediction Class Index: 207
Inference Time: 6383.15 ms
Size (MB): 20.32 MB
Parameters: 5326716
Depth: 771

```
[45]: import numpy as np
from tensorflow.keras.applications import NASNetLarge
from tensorflow.keras.preprocessing import image
from tensorflow.keras.applications.nasnet import preprocess_input, decode_predictions

# Load the NASNetLarge model pre-trained on ImageNet data
model = NASNetLarge(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the path to your image file
img = image.load_img(img_path, target_size=(331, 331)) # NASNetLarge requires input shape (331, 331)
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
```

```

start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")

```

1/1 [=====] - 9s 9s/step

Predictions:

- 1: golden_retriever (0.91)
- 2: Labrador_retriever (0.01)
- 3: tennis_ball (0.00)
- 4: Airedale (0.00)
- 5: Irish_setter (0.00)

Top Prediction Class Index: 207

Inference Time: 9299.89 ms

Size (MB): 339.32 MB

Parameters: 88949818

Depth: 1041

```
[47]: import numpy as np
from tensorflow.keras.applications import EfficientNetV2B0
from tensorflow.keras.preprocessing import image
from tensorflow.keras.applications.efficientnet_v2 import preprocess_input, u
↳decode_predictions

# Load the EfficientNetV2B0 model pre-trained on ImageNet data
model = EfficientNetV2B0(weights='imagenet')

# Load and preprocess the input image
img_path = r"C:\Users\srava\Downloads\Nassau_2697x1517.jpg" # replace with the u
↳path to your image file
img = image.load_img(img_path, target_size=(224, 224)) # NASNetLarge requires u
↳input shape (331, 331)
img_array = image.img_to_array(img)
img_array = np.expand_dims(img_array, axis=0)
img_array = preprocess_input(img_array)

# Measure inference time
start_time = time.time()
predictions = model.predict(img_array)
end_time = time.time()

# Decode and print the top-3 predicted classes
decoded_predictions = decode_predictions(predictions, top=5)[0]

print("Predictions:")
for i, (imagenet_id, label, score) in enumerate(decoded_predictions):
    print(f"{i + 1}: {label} ({score:.2f})")

# Optionally, you can obtain the class index for the top prediction
top_class_index = np.argmax(predictions[0])
print(f"\nTop Prediction Class Index: {top_class_index}")

# Calculate and print the inference time per step
inference_time_ms = (end_time - start_time) * 1000.0
print(f"Inference Time: {inference_time_ms:.2f} ms")

# Model summary provides information about parameters and layers
#model.summary()

# Get the size of the model in megabytes
model_size_MB = model.count_params() * 4 / (1024 ** 2) # 4 bytes for float32
print(f"Size (MB): {model_size_MB:.2f} MB")

# Get the number of parameters and depth from the model's layers
```

```
num_parameters = model.count_params()
model_depth = len(model.layers)

print(f"Parameters: {num_parameters}")
print(f"Depth: {model_depth}")
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-

applications/efficientnet_v2/efficientnetv2-b0.h5

29403144/29403144 [=====] - 11s 0us/step

1/1 [=====] - 3s 3s/step

Predictions:

1: golden_retriever (0.92)
2: Labrador_retriever (0.01)
3: Tibetan_mastiff (0.00)
4: cocker_spaniel (0.00)
5: kuvasz (0.00)

Top Prediction Class Index: 207

Inference Time: 2635.39 ms

Size (MB): 27.47 MB

Parameters: 7200312

Depth: 273

[]: