## **Tools Setup**

```
In [ ]: import pathlib, os, sys, operator, re, datetime
        from functools import reduce
        import numpy as np
        import tensorflow as tf
        import numpy as np
        import tensorflow as tf
        import matplotlib.pyplot as plt
        from tensorflow import keras
        from keras import Model
        import tensorflow datasets as tfds
        from tiny imagenet import TinyImagenetDataset
        # Enable or disable GPU
        # To fully disable it, we need to hide all GPU devices from Tensorflow
        # Make sure GPU is disabled for this inference part of the lab
        ENABLE GPU = True
        # tf.debugging.set log device placement(True)
        if not ENABLE GPU:
            tf.config.set visible devices([], 'GPU')
        # Print Python and TF version, and where we are running
        print(f'Running on Python Version: {sys.version}')
        print(f'Using Tensorflow Version: {tf. version }')
        if not tf.config.experimental.list physical devices("GPU"):
            print('Running on CPU')
        else:
            print(f'Using GPU at: {tf.test.gpu device name()} (of {len(tf.config.exp
```

2025-09-08 12:07:10.459501: I tensorflow/core/util/port.cc:113] oneDNN custo m operations are on. You may see slightly different numerical results due to floating-point round-off errors from different computation orders. To turn t hem off, set the environment variable `TF ENABLE ONEDNN OPTS=0`. 2025-09-08 12:07:11.461606: E external/local xla/xla/stream executor/cuda/cu da fft.cc:479] Unable to register cuFFT factory: Attempting to register fact ory for plugin cuFFT when one has already been registered 2025-09-08 12:07:11.941786: E external/local xla/xla/stream executor/cuda/cu da dnn.cc:10575] Unable to register cuDNN factory: Attempting to register fa ctory for plugin cuDNN when one has already been registered 2025-09-08 12:07:11.945026: E external/local xla/xla/stream executor/cuda/cu da blas.cc:1442] Unable to register cuBLAS factory: Attempting to register f actory for plugin cuBLAS when one has already been registered 2025-09-08 12:07:12.678895: I tensorflow/core/platform/cpu feature guard.cc: 210] This TensorFlow binary is optimized to use available CPU instructions i n performance-critical operations. To enable the following instructions: AVX2 AVX VNNI FMA, in other operation s, rebuild TensorFlow with the appropriate compiler flags. 2025-09-08 12:07:16.219884: W tensorflow/compiler/tf2tensorrt/utils/py util s.cc:38] TF-TRT Warning: Could not find TensorRT

```
Running on Python Version: 3.9.21 (main, Feb 10 2025, 00:00:00)
[GCC 11.5.0 20240719 (Red Hat 11.5.0-5)]
Using Tensorflow Version: 2.16.2
Running on CPU
2025-09-08 12:07:25.898466: I external/local xla/xla/stream executor/cuda/cu
da diagnostics.cc:134] retrieving CUDA diagnostic information for host: co20
50-19.ece.iastate.edu
2025-09-08 12:07:25.898489: I external/local xla/xla/stream executor/cuda/cu
da diagnostics.cc:141] hostname: co2050-19.ece.iastate.edu
2025-09-08 12:07:25.898551: I external/local xla/xla/stream executor/cuda/cu
da diagnostics.cc:165] libcuda reported version is: NOT FOUND: was unable to
find libcuda.so DSO loaded into this program
2025-09-08 12:07:25.898577: I external/local xla/xla/stream executor/cuda/cu
da diagnostics.cc:169] kernel reported version is: NOT FOUND: could not find
kernel module information in driver version file contents: "NVRM version: NV
IDIA UNIX Open Kernel Module for x86 64 580.65.06 Release Build (root@co2
050-19.ece.iastate.edu) Fri Aug 15 15:09:32 CDT 2025
GCC version: gcc version 11.5.0 20240719 (Red Hat 11.5.0-5) (GCC)
```

## **Dataset Inspection**

```
In [2]: # This cell imports our dataset.
        # Original Source: https://github.com/ksachdeva/tiny-imagenet-tfds
        # Class Version Source: https://github.com/duweisu/tiny-imagenet-tfds
        # Setup our dataset
        tiny imagenet builder = TinyImagenetDataset()
        # this call (download and prepare) will trigger the download of the dataset
        # and preparation (conversion to tfrecords)
        # This will be done only once and on next usage tfds will
        # use the cached version on your host.
        tiny imagenet builder.download and prepare(download dir="~/tensorflow-datase
        # class names = tiny imagenet builder.info.features['label'].names
        ds = tiny imagenet builder.as dataset()
        ds train, ds val = ds["train"], ds["validation"]
        assert(isinstance(ds train, tf.data.Dataset))
        assert(isinstance(ds val, tf.data.Dataset))
        # Training Dataset
        ds train = ds train.shuffle(1024).prefetch(tf.data.AUTOTUNE)
        # Validation Dataset
        ds val = ds val.shuffle(1024).prefetch(tf.data.AUTOTUNE)
        # Dataset metadata
        ds info = tiny imagenet builder.info
```

WARNING:absl:You use TensorFlow DType <dtype: 'string'> in tfds.features Thi s will soon be deprecated in favor of NumPy DTypes. In the meantime it was c onverted to object.

```
In [3]: # We need to read the "human readable" labels so we can translate with the r
        # Read the labels file (words.txt)
        with open(os.path.abspath('wnids.txt'), 'r') as f:
            wnids = [x.strip() for x in f]
        # Map wnids to integer labels
        wnid to label = {wnid: i for i, wnid in enumerate(wnids)}
        label to wnid = {v: k for k, v in wnid to label.items()}
        # Use words.txt to get names for each class
        with open(os.path.abspath('words.txt'), 'r') as f:
            wnid to words = dict(line.split('\t') for line in f)
            for wnid, words in wnid to words.items():
                wnid to words[wnid] = [w.strip() for w in words.split(',')]
        class names = [str(wnid to words[wnid]) for wnid in wnids]
In [4]: # Helper function to get the label name
        def img class(img data, idx=None):
            image, label, id, label name = img data["image"], img data["label"], imc
            # Handle batches of images correctly
            if idx != None:
                image, label, id, label name = img data["image"][idx], img data["lat
            return f"{label name} (class index: {label} - id: {id})"
        # Helper function to show basic info about an image
        def img info(img, idx=None, display=True, title apend=""):
            image = img['image']
            # Print the class
            class str = img class(img, idx)
            print(f"Label: {class str}")
            # Display the image
            if display:
                plt.figure()
                plt.title(title apend + class str)
                # Handle batches correctly
                if image.shape.ndims > 3:
                    plt.imshow(image.numpy().reshape(64, 64, 3))
                else:
                    plt.imshow(image.numpy())
```

```
In [5]: # Print the dataset types and info
    print("--- Train & Validation dataset info ---")
    print(f"Train: {ds_train}")
    print(f"Validation: {ds_val}")
    # print(f"Dataset Info: {ds_info}") # Uncomment to print Dataset info
```

```
print("\n--- Show an example image ---")
for example in ds_val.take(1):
    img_info(example)

print("\n Show some other examples")
tfds.show_examples(ds_val, ds_info, rows=3, cols=3)
```

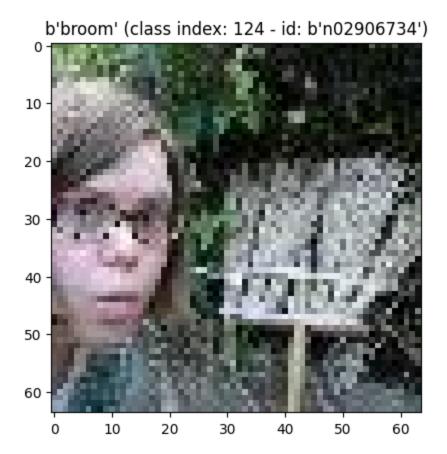
--- Train & Validation dataset info ---

Train: <\_PrefetchDataset element\_spec={'id': TensorSpec(shape=(), dtype=tf.s tring, name=None), 'image': TensorSpec(shape=(64, 64, 3), dtype=tf.uint8, na me=None), 'label': TensorSpec(shape=(), dtype=tf.int64, name=None), 'metadat a': {'label\_name': TensorSpec(shape=(), dtype=tf.string, name=None)}}> Validation: <\_PrefetchDataset element\_spec={'id': TensorSpec(shape=(), dtype=tf.string, name=None), 'image': TensorSpec(shape=(64, 64, 3), dtype=tf.uint 8, name=None), 'label': TensorSpec(shape=(), dtype=tf.int64, name=None), 'me tadata': {'label name': TensorSpec(shape=(), dtype=tf.string, name=None)}}>

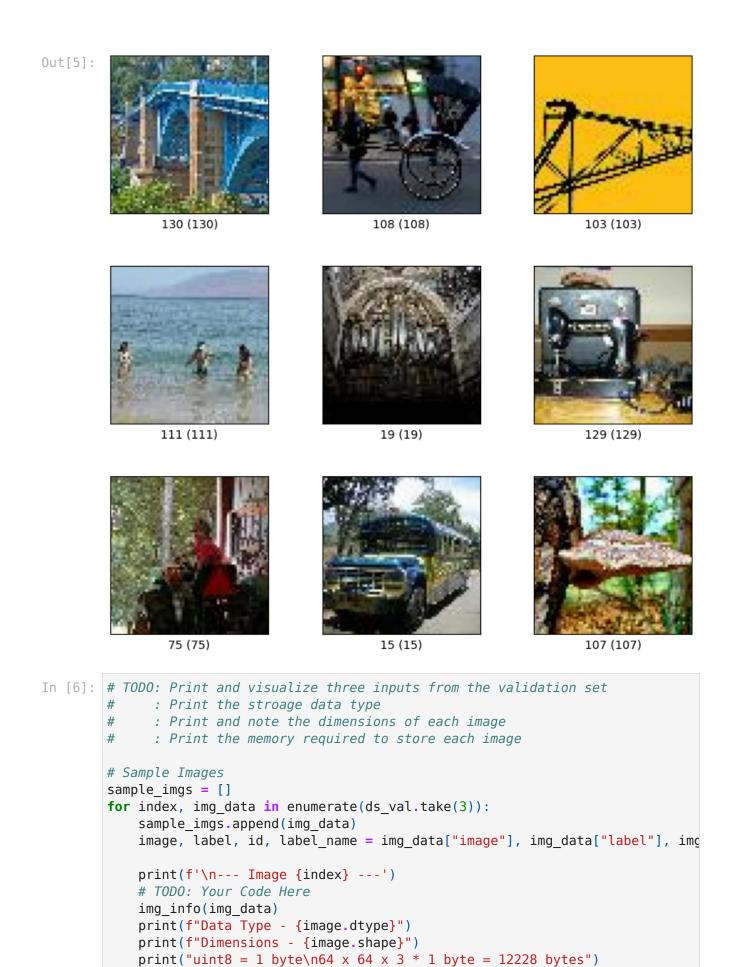
--- Show an example image ---

2025-09-08 12:07:26.349885: W tensorflow/core/kernels/data/cache dataset op s.cc:858] The calling iterator did not fully read the dataset being cached. In order to avoid unexpected truncation of the dataset, the partially cached contents of the dataset will be discarded. This can happen if you have an i nput pipeline similar to `dataset.cache().take(k).repeat()`. You should use `dataset.take(k).cache().repeat()` instead. 2025-09-08 12:07:26.401154: W tensorflow/core/framework/local rendezvous.cc: 404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence 2025-09-08 12:07:26.492776: W tensorflow/core/kernels/data/cache dataset op s.cc:858] The calling iterator did not fully read the dataset being cached. In order to avoid unexpected truncation of the dataset, the partially cached contents of the dataset will be discarded. This can happen if you have an i nput pipeline similar to `dataset.cache().take(k).repeat()`. You should use `dataset.take(k).cache().repeat()` instead. 2025-09-08 12:07:26.506813: W tensorflow/core/framework/local rendezvous.cc: 404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence Label: b'broom' (class index: 124 - id: b'n02906734')

Show some other examples







# See example usage: https://github.com/duweisu/tiny-imagenet-tfds

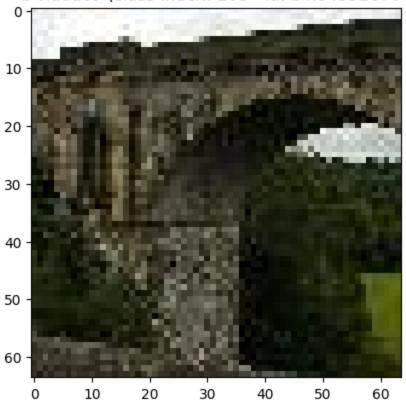
```
--- Image 0 ---
Label: b'viaduct' (class index: 168 - id: b'n04532670')
Data Type - <dtype: 'uint8'>
Dimensions - (64, 64, 3)
uint8 = 1 byte
64 \times 64 \times 3 * 1 \text{ byte} = 12228 \text{ bytes}
--- Image 1 ---
Label: b'iPod' (class index: 91 - id: b'n03584254')
Data Type - <dtype: 'uint8'>
Dimensions - (64, 64, 3)
uint8 = 1 byte
64 \times 64 \times 3 * 1 \text{ byte} = 12228 \text{ bytes}
--- Image 2 ---
Label: b'pretzel' (class index: 40 - id: b'n07695742')
Data Type - <dtype: 'uint8'>
Dimensions - (64, 64, 3)
uint8 = 1 byte
64 \times 64 \times 3 * 1 \text{ byte} = 12228 \text{ bytes}
2025-09-08 12:07:27.232384: W tensorflow/core/kernels/data/cache dataset op
s.cc:858] The calling iterator did not fully read the dataset being cached.
In order to avoid unexpected truncation of the dataset, the partially cached
contents of the dataset will be discarded. This can happen if you have an i
```

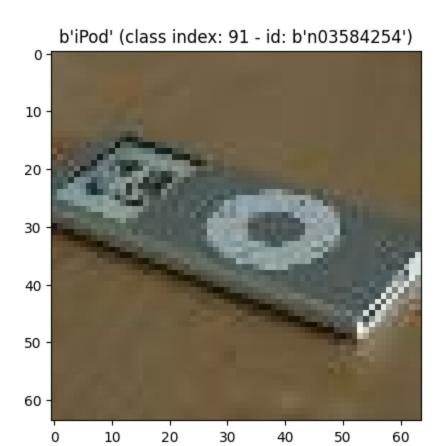
nput pipeline similar to `dataset.cache().take(k).repeat()`. You should use

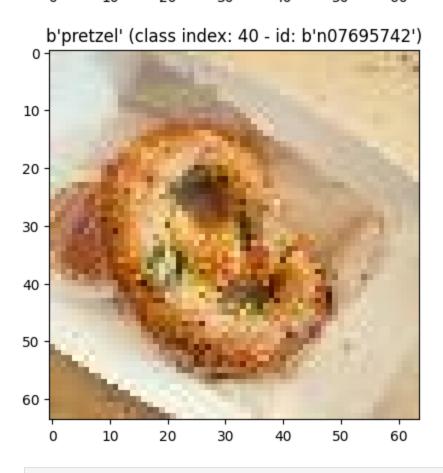
2025-09-08 12:07:27.238809: W tensorflow/core/framework/local\_rendezvous.cc: 404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence

### b'viaduct' (class index: 168 - id: b'n04532670')

`dataset.take(k).cache().repeat()` instead.







In [7]: # TODO: Export each of the three inputs to a binary file which will be used # NOTE: First flatten the array (ex: 4D --> 1D). So 64\*64\*3 = 12288 element

```
# Make a directory for our image data
img dir = os.path.abspath('img data')
pathlib.Path(img dir).mkdir(exist ok=True)
# Create a metadata file
metadata file = open(os.path.join(img dir, f'metadata.txt'), 'w')
metadata file.write(f'Number\t\tDims\t\tClass Data\n')
# Export each image
for index, img data in enumerate(sample imgs):
    img file = open(os.path.join(img dir, f'image {index}.bin'), 'wb')
    # TODO: Your Code Here
    arr = img data["image"].numpy()
    label = img data["label"].numpy()
    flat arr = arr.flatten()
    img file.write(flat arr.tobytes())
    img file.close()
    metadata file.write(f"{index}\t\t{image.shape}\t\t{label}\n")
```

# Model Loading and Inference

```
In [8]: # TODO: Load the model
        # Now we will load the H5 model! Please make sure the h5 model file is prese
        # You can download this from the Canvas Page and place it in the same direct
        # model path = os.path.abspath(""/home/<NETID>/path/to/your/lab1/CNN TinyIma
        model path = os.path.abspath("CNN TinyImageNet.h5")
        # TODO: Your Code Here
        model = tf.keras.models.load model(model path)
        model.summary()
       /home/buenting/cpre 5870/lab1/Lab1/lab1 venv/lib64/python3.9/site-packages/k
       eras/src/layers/convolutional/base conv.py:113: 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 instea
```

d.

super(). init (activity regularizer=activity regularizer, \*\*kwargs) /home/buenting/cpre 5870/lab1/Lab1/lab1 venv/lib64/python3.9/site-packages/k eras/src/optimizers/base optimizer.py:86: UserWarning: Argument `decay` is n o longer supported and will be ignored.

warnings.warn(

WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile metrics` will be empty until you train or evaluate the model.

WARNING:absl:Error in loading the saved optimizer state. As a result, your m odel is starting with a freshly initialized optimizer.

Model: "sequential"

Layer (type)	Output Shape	Par
conv2d (Conv2D)	(None, 60, 60, 32)	2
conv2d_1 (Conv2D)	(None, 56, 56, 32)	25
max_pooling2d (MaxPooling2D)	(None, 28, 28, 32)	
conv2d_2 (Conv2D)	(None, 26, 26, 64)	18
conv2d_3 (Conv2D)	(None, 24, 24, 64)	36
max_pooling2d_1 (MaxPooling2D)	(None, 12, 12, 64)	
conv2d_4 (Conv2D)	(None, 10, 10, 64)	36
conv2d_5 (Conv2D)	(None, 8, 8, 128)	73
max_pooling2d_2 (MaxPooling2D)	(None, 4, 4, 128)	
flatten (Flatten)	(None, 2048)	
dense (Dense)	(None, 256)	524
dense_1 (Dense)	(None, 200)	51

Total params: 770,218 (2.94 MB)

Trainable params: 770,216 (2.94 MB)

Non-trainable params: 0 (0.00 B)

Optimizer params: 2 (12.00 B)

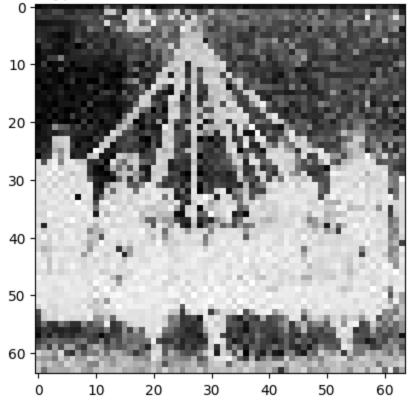
```
In [9]: # Running infrence on our model
# We can run an infrence of our model by doing the following (we are doing the for example in ds_train.batch(1).take(1):
    img_info(example)
    # Make a prediction
    pred = model.predict(tf.cast(example["image"], tf.float32)/255.0)
    # print(f'Raw 200 Class Weighted Prediction:\n{pred}') # Uncomment to se

# What is out best guess?
    best_guess = tf.math.argmax(pred, axis=1).numpy() # Our output is 200 we print(f'Best Guess [class index]: {class_names[best_guess[0]]} [{best_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_guest_g
```

2025-09-08 12:07:27.843166: W tensorflow/core/kernels/data/cache\_dataset\_op s.cc:858] The calling iterator did not fully read the dataset being cached. In order to avoid unexpected truncation of the dataset, the partially cached contents of the dataset will be discarded. This can happen if you have an input pipeline similar to `dataset.cache().take(k).repeat()`. You should use `dataset.take(k).cache().repeat()` instead.

```
Label: [b'maypole'] (class index: [21] - id: [b'n03733131'])
                         — 0s 264ms/step
Best Guess [class index]: ['tabby', 'tabby cat'] [66]
Best Guess Confidence (percent / 1.0): [0.12290254]
Top 15 Guesses (class index): ["['tabby', 'tabby cat'] [66]", "['nail'] [2 9]", "['walking stick', 'walkingstick', 'stick insect'] [181]", "['fountai
n'] [140]", "['suspension bridge'] [119]", "['stopwatch', 'stop watch'] [13
7]", "['American alligator', 'Alligator mississipiensis'] [160]", "['steel a
rch bridge'] [130]", "['chain'] [171]", "['rocking chair', 'rocker'] [3]", "
['koala', 'koala bear', 'kangaroo bear', 'native bear', 'Phascolarctos ciner
eus'] [68]", "['magnetic compass'] [191]", "['maypole'] [21]", "['birdhous
e'] [72]", "['Egyptian cat'] [0]"]
Top 15 Guesses Confidence (percent / 1.0): [[0.12290254 0.09233914 0.0604626
7 0.05863593 0.05133322 0.04556591
  0.0418246 \quad 0.03907003 \quad 0.03407361 \quad 0.02940531 \quad 0.02923596 \quad 0.0203824
  0.0202339 0.01993833 0.01878766]]
2025-09-08 12:07:28.178480: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
```

### [b'maypole'] (class index: [21] - id: [b'n03733131'])



```
In [10]: # TODO: Run infrence for our previous 3 sample images

# TODO: Your Code Here

# for index, img_data in enumerate(sample_imgs):

# pred = model.predict(tf.cast(img_data["image"], tf.float32)/255.0)

# sample_predictions.append(pred)

# best_guess = tf.math.argmax(sample_predictions[index], axis=1).numpy()

# label = sample_predictions[index]["label"].numpy()

# print(f"{best_guess}\t\t{pred[0][best_guess]}\t\t\t")
```

```
# Stack all 3 sample images into a stack for the prediction model (3, 64, 64
        imgs = tf.stack([tf.cast(d["image"], tf.float32) / 255.0 for d in sample imc
        preds = model.predict(imgs)
        print(f"Best guess\t\tAssoc. Conf.\t\tActual Class\n")
        for i, pred in enumerate(preds):
            best guess = np.argmax(pred)
            confidence = pred[best quess]
            label = sample imgs[i]["label"].numpy()
            print(f"{best quess}\t\t{confidence}\t\t{label}")
       1/1 -
                               • 0s 94ms/step
       Best quess
                               Assoc. Conf.
                                                       Actual Class
       168
                       0.4106539785861969
                                                       168
       26
                                                       91
                       0.2709677219390869
       40
                       0.4427526593208313
                                                       40
In [ ]: | # # TODO: Calculate the Top-1, Top-5, and Top-10 Accuracy of the validation
        total = acc top1 = acc top5 = acc top10 = 0
        total = ds val.cardinality().numpy()
        # # TODO: Your Code Here
        for images in ds val.batch(32):
            # Predict the classes for each image, in a batch of 32 images at a time
            preds = model.predict(tf.cast(images["image"], tf.float32)/255.0, verbos
            labels = images["label"].numpy()
            for i, pred in enumerate(preds):
                top 1 = np.argmax(pred)
                if top 1 == labels[i]:
                    acc top1 += 1
                # top 5 = tf.math.top k(pred, n=5)
                # for j in range(5):
                # if top 5[i] == labels[i]:
                #
                         acc top5 += 1
                #
                         break
                # top 10 = tf.math.top k(pred, n=10)
                # for j in range(10):
                # if top 10[i] == labels[i]:
                #
                        acc top10 += 1
                #
                         break
                top 5 = tf.math.top k(pred, k=5).indices.numpy()
                if labels[i] in top 5:
                    acc top5 += 1
                top_10 = tf.math.top_k(pred, k=10).indices.numpy()
                if labels[i] in top 10:
                    acc top10 += 1
        print(f"Top 1 Accuracy: {acc top1/total}\n")
```

```
print(f"Top 5 Accuracy: {acc top5/total}\n")
         print(f"Top 10 Accuracy: {acc top10/total}\n")
        Top 1 Accuracy: 0.243
        Top 5 Accuracy: 0.494
        Top 10 Accuracy: 0.6174
        2025-09-08 12:07:45.973865: W tensorflow/core/framework/local rendezvous.cc:
        404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
In [12]: # # TODO: Print all of the possible classes of the dataset
         # # TODO: Your Code Here
         train labels = set()
         for batch in ds train.batch(32):
             train labels.update(batch["label"].numpy().tolist())
         val labels = set()
         for batch in ds train.batch(32):
             val labels.update(batch["label"].numpy().tolist())
         print("Training classes:", len(train labels))
         print("Validation classes:", len(val labels))
         for i in range(200):
             print({class names[i],})
        2025-09-08 12:07:48.673151: W tensorflow/core/framework/local rendezvous.cc:
        404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
```

```
Training classes: 200
Validation classes: 200
{"['Egyptian cat']"}
{"['reel']"}
{"['volleyball']"}
{"['rocking chair', 'rocker']"}
{"['lemon']"}
{"['bullfrog', 'Rana catesbeiana']"}
{"['basketball']"}
{"['cliff', 'drop', 'drop-off']"}
{"['espresso']"}
{'[\'plunger\', "plumber\'s helper"]'}
{"['parking meter']"}
{"['German shepherd', 'German shepherd dog', 'German police dog', 'alsatia
n']"}
{"['dining table', 'board']"}
{"['monarch', 'monarch butterfly', 'milkweed butterfly', 'Danaus plexippu
s']"}
{"['brown bear', 'bruin', 'Ursus arctos']"}
{"['school bus']"}
{"['pizza', 'pizza pie']"}
{"['guinea pig', 'Cavia cobaya']"}
{"['umbrella']"}
{"['organ', 'pipe organ']"}
{"['oboe', 'hautboy', 'hautbois']"}
{"['maypole']"}
{"['goldfish', 'Carassius auratus']"}
{"['potpie']"}
{"['hourglass']"}
{"['seashore', 'coast', 'seacoast', 'sea-coast']"}
{"['computer keyboard', 'keypad']"}
{"['Arabian camel', 'dromedary', 'Camelus dromedarius']"}
{"['ice cream', 'icecream']"}
{"['nail']"}
{"['space heater']"}
{"['cardigan']"}
{"['baboon']"}
{"['snail']"}
{"['coral reef']"}
{"['albatross', 'mollymawk']"}
{'[\'spider web\', "spider\'s web"]'}
{"['sea cucumber', 'holothurian']"}
{"['backpack', 'back pack', 'knapsack', 'packsack', 'rucksack', 'haversac
k']"}
{"['Labrador retriever']"}
{"['pretzel']"}
{"['king penguin', 'Aptenodytes patagonica']"}
{"['sulphur butterfly', 'sulfur butterfly']"}
{"['tarantula']"}
{"['lesser panda', 'red panda', 'panda', 'bear cat', 'cat bear', 'Ailurus fu
lgens']"}
{"['pop bottle', 'soda bottle']"}
{"['banana']"}
{"['sock']"}
{"['cockroach', 'roach']"}
{"['projectile', 'missile']"}
```

```
{"['beer bottle']"}
{"['mantis', 'mantid']"}
{"['freight car']"}
{"['quacamole']"}
{"['remote control', 'remote']"}
{"['European fire salamander', 'Salamandra salamandra']"}
{"['lakeside', 'lakeshore']"}
{"['chimpanzee', 'chimp', 'Pan troglodytes']"}
{"['pay-phone', 'pay-station']"}
{"['fur coat']"}
{"['alp']"}
{"['lampshade', 'lamp shade']"}
{"['torch']"}
{"['abacus']"}
{"['moving van']"}
{"['barrel', 'cask']"}
{"['tabby', 'tabby cat']"}
{"['goose']"}
{"['koala', 'koala bear', 'kangaroo bear', 'native bear', 'Phascolarctos cin
ereus']"}
{"['bullet train', 'bullet']"}
{"['CD player']"}
{"['teapot']"}
{"['birdhouse']"}
{"['gazelle']"}
{'[\'academic gown\', \'academic robe\', "judge\'s robe"]'}
{"['tractor']"}
{"['ladybug', 'ladybeetle', 'lady beetle', 'ladybird', 'ladybird beetle']"}
{"['miniskirt', 'mini']"}
{"['golden retriever']"}
{"['triumphal arch']"}
{"['cannon']"}
{"['neck brace']"}
{"['sombrero']"}
{"['gasmask', 'respirator', 'gas helmet']"}
{"['candle', 'taper', 'wax light']"}
{"['desk']"}
{"['frying pan', 'frypan', 'skillet']"}
{"['bee']"}
{"['dam', 'dike', 'dyke']"}
{"['spiny lobster', 'langouste', 'rock lobster', 'crawfish', 'crayfish', 'se
a crawfish']"}
{"['police van', 'police wagon', 'paddy wagon', 'patrol wagon', 'wagon', 'bl
ack Maria']"}
{"['iPod']"}
{"['punching bag', 'punch bag', 'punching ball', 'punchball']"}
{"['beacon', 'lighthouse', 'beacon light', 'pharos']"}
{"['jellyfish']"}
{"['wok']"}
{'["potter\'s wheel"]'}
{"['sandal']"}
{"['pill bottle']"}
{"['butcher shop', 'meat market']"}
{"['slug']"}
{"['hog', 'pig', 'grunter', 'squealer', 'Sus scrofa']"}
{"['cougar', 'puma', 'catamount', 'mountain lion', 'painter', 'panther', 'Fe
```

```
lis concolor']"}
{"['crane']"}
{"['vestment']"}
{'[\'dragonfly\', \'darning needle\', "devil\'s darning needle", \'sewing ne
edle\', \'snake feeder\', \'snake doctor\', \'mosquito hawk\', \'skeeter haw
k\']'}
{"['cash machine', 'cash dispenser', 'automated teller machine', 'automatic
teller machine', 'automated teller', 'automatic teller', 'ATM']"}
{"['mushroom']"}
{"['jinrikisha', 'ricksha', 'rickshaw']"}
{"['water tower']"}
{"['chest']"}
{"['snorkel']"}
{"['sunglasses', 'dark glasses', 'shades']"}
{"['fly']"}
{"['limousine', 'limo']"}
{"['black stork', 'Ciconia nigra']"}
{"['dugong', 'Dugong dugon']"}
{"['sports car', 'sport car']"}
{"['water jug']"}
{"['suspension bridge']"}
{"['ox']"}
{"['ice lolly', 'lolly', 'lollipop', 'popsicle']"}
{"['turnstile']"}
{"['Christmas stocking']"}
{"['broom']"}
{"['scorpion']"}
{"['wooden spoon']"}
{"['picket fence', 'paling']"}
{"['rugby ball']"}
{"['sewing machine']"}
{"['steel arch bridge']"}
{"['Persian cat']"}
{"['refrigerator', 'icebox']"}
{"['barn']"}
{"['apron']"}
{"['Yorkshire terrier']"}
{"['swimming trunks', 'bathing trunks']"}
{"['stopwatch', 'stop watch']"}
{"['lawn mower', 'mower']"}
{"['thatch', 'thatched roof']"}
{"['fountain']"}
{"['black widow', 'Latrodectus mactans']"}
{"['bikini', 'two-piece']"}
{"['plate']"}
{"['teddy', 'teddy bear']"}
{"['barbershop']"}
\{"['confectionery', 'confectionary', 'candy store']"\}
{"['beach wagon', 'station wagon', 'wagon', 'estate car', 'beach waggon', 's
tation waggon', 'waggon']"}
{"['scoreboard']"}
{"['orange']"}
{"['flagpole', 'flagstaff']"}
{"['American lobster', 'Northern lobster', 'Maine lobster', 'Homarus america
{"['trolleybus', 'trolley coach', 'trackless trolley']"}
```

```
{"['drumstick']"}
{"['dumbbell']"}
{"['brass', 'memorial tablet', 'plaque']"}
{"['bow tie', 'bow-tie', 'bowtie']"}
{"['convertible']"}
{"['bighorn', 'bighorn sheep', 'cimarron', 'Rocky Mountain bighorn', 'Rocky
Mountain sheep', 'Ovis canadensis']"}
{"['orangutan', 'orang', 'orangutang', 'Pongo pygmaeus']"}
{"['American alligator', 'Alligator mississipiensis']"}
{"['centipede']"}
{"['syringe']"}
{"['go-kart']"}
{"['brain coral']"}
{"['sea slug', 'nudibranch']"}
{"['cliff dwelling']"}
{"['mashed potato']"}
{"['viaduct']"}
{"['military uniform']"}
{"['pomegranate']"}
{"['chain']"}
{"['kimono']"}
{"['comic book']"}
{"['trilobite']"}
{"['bison']"}
{"['pole']"}
{"['boa constrictor', 'Constrictor constrictor']"}
{"['poncho']"}
{"['bathtub', 'bathing tub', 'bath', 'tub']"}
{"['grasshopper', 'hopper']"}
{"['walking stick', 'walkingstick', 'stick insect']"}
{"['Chihuahua']"}
{"['tailed frog', 'bell toad', 'ribbed toad', 'tailed toad', 'Ascaphus tru
i']"}
{"['lion', 'king of beasts', 'Panthera leo']"}
{"['altar']"}
{"['obelisk']"}
{"['beaker']"}
{"['bell pepper']"}
{"['bannister', 'banister', 'balustrade', 'balusters', 'handrail']"}
{"['bucket', 'pail']"}
{"['magnetic compass']"}
{"['meat loaf', 'meatloaf']"}
{"['gondola']"}
{"['standard poodle']"}
{"['acorn']"}
{"['lifeboat']"}
{"['binoculars', 'field glasses', 'opera glasses']"}
{"['cauliflower']"}
{"['African elephant', 'Loxodonta africana']"}
2025-09-08 12:07:48.944399: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
```

## **Model Exploration**

```
In [13]: # TODO: Visualize the model in Netron (https://netron.app/) and include an i
         # tf.keras.utils.plot model(model, "model.png", show shapes=True, show dtype
In [14]: # We can view the layer weights. Here we consider them as images of feature
         # TODO: Visualize the 2 convolutional layers filter sets (weights) (one at t
         # TODO: Your Code Here
         conv layers = [l for l in model.layers if isinstance(l, tf.keras.layers.Conv
         first conv = conv layers[0]
         last conv = conv layers[-2]
         def describe layer(layer):
             W, b = layer.get weights()
             n params = W.size + b.size
             mem bytes = W.nbytes + b.nbytes
             print(f"\nLayer: {layer.name} ({layer.__class__.__name__})")
             print(f"Weights shape: {W.shape}, Bias shape: {b.shape}")
             print(f"dtype: {W.dtype}")
             print(f"Params: {n params:,}")
             print(f"Memory: {mem bytes} bytes")
             return W,b
         W1, b1 = describe layer(first conv)
         W2, b2 = describe layer(last conv)
         def normalize(x):
             x = x - x.min()
             m = x.min()
             return x / (m if m != 0 else 1.0)
         num show1 = min(16, W1.shape[-1])
         cols = 8
         rows = int(num show1 / cols)
         plt.figure()
         for i in range(num show1):
             f = W1[:, :, :, i]
             f img = normalize(f)
             ax = plt.subplot(rows, cols, i+1)
             ax.imshow(f_img)
             ax.set xticks([]); ax.set yticks([])
         plt.suptitle(f"{first conv.name} first {num show1} filters")
         plt.show()
         num show2 = min(16, W2.shape[-1])
         rows = int(num_show2 / cols)
         plt.figure()
         for i in range(num show2):
             k = W2[:, :, :, i]
             energy = np.sqrt((k**2).sum(axis=2))
             e img = normalize(energy)
             ax = plt.subplot(rows, cols, i+1)
             ax.imshow(e img)
             ax.set xticks([]); ax.set yticks([])
         plt.suptitle(f"{last conv.name} first {num show2} filters")
         plt.show()
```

Layer: conv2d (Conv2D)

Weights shape: (5, 5, 3, 32), Bias shape: (32,)

dtype: float32
Params: 2,432
Memory: 9728 bytes

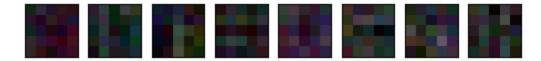
Layer: conv2d\_4 (Conv2D)

Weights shape: (3, 3, 64, 64), Bias shape: (64,)

dtype: float32
Params: 36,928
Memory: 147712 bytes

conv2d first 16 filters





conv2d\_4 first 16 filters



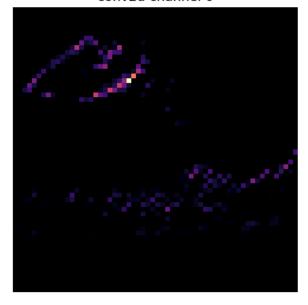


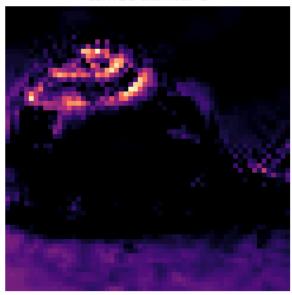
```
In [15]: # We can view the layer outputs as well. Here we consider them as images of
         # TODO: Visualize the 2 convolutional layers outputs (intermediate feature n
         # TODO: Your Code Here
         example = next(iter(ds val.batch(1)))
         x = tf.cast(example["image"], tf.float32) / 255.0
         acts = {}
         t = x
         for layer in model.layers:
             try:
                 t = layer(t, training = False)
             except TypeError:
                 t = layer(t)
             if layer is first conv:
                 acts["first"] = t.numpy()
             if layer is last conv:
                 acts["last"] = t.numpy()
         feat first = acts["first"]
         feat last = acts["last"]
         def show feature maps(F, layer name, channels=(0, -1), cmap = "magma"):
             F0 = np.asarray(F)[0]
             H, W, C = F0.shape
             print(f"Layer name: {layer name} ")
             print(f"feature map shape = {(1, H, W, C)}")
             print(f"dtype{F.dtype}")
             print(f"Data size: {F.nbytes}")
             plt.figure(figsize=(4*len(channels), 4))
             for i, ch in enumerate(channels):
                 fm = F0[..., ch]
                 fm = (fm - fm.min()) / (fm.max() - fm.min() + 1e-8)
                 ax = plt.subplot(1, len(channels), i+1)
                 ax.imshow(fm, cmap=cmap)
                 ax.set title(f"{layer name} channel {ch}")
                 ax.axis("off")
             plt.tight layout()
             plt.show()
         show feature maps(feat first, first conv.name)
         show feature maps(feat last, last conv.name)
```

Layer name: conv2d feature map shape = (1, 60, 60, 32) dtypefloat32 Data size: 460800

#### conv2d channel 0

#### conv2d channel -1



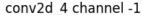


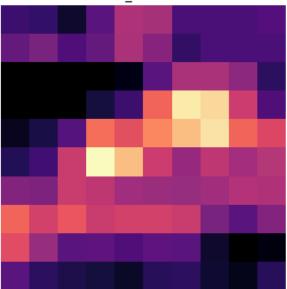
Layer name: conv2d 4

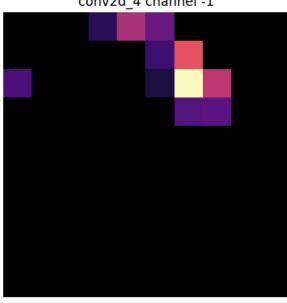
feature map shape = (1, 10, 10, 64)

dtypefloat32 Data size: 25600

conv2d 4 channel 0







```
In [16]: # TODO: Export the filters/weights se we can use them later
         # Make a directory for our image data
         model dir = os.path.abspath('model data')
         pathlib.Path(model_dir).mkdir(exist_ok=True)
         # Export each image
         conv index = dense index = 1 # layer index starts from one
         for layer idx, layer in enumerate(model.layers):
             if re.match(r'(conv|dense)', layer.name):
                 weight file name = os.path.join(model dir, f'{layer.name} weights.bi
                 bias file name = os.path.join(model dir, f'{layer.name} bias.bin')
             else: continue
```

```
assert layer.weights[0].name.endswith('kernel')
assert layer.weights[1].name.endswith('bias')

# TODO: Your Code Here
W, b = layer.get_weights()

with open(weight_file_name, "wb") as wf:
    wf.write(W.tobytes())

with open(bias_file_name, "wb") as bf:
    bf.write(b.tobytes())
```

```
In [17]: # TODO: Export the intermediate layer outputs for each of the input for all
         img dir = os.path.abspath('img data')
         pathlib.Path(img dir).mkdir(exist ok=True)
         for img idx, img in enumerate(sample imgs):
             file dir = os.path.join(img dir, f'test input {img idx}')
             pathlib.Path(file dir).mkdir(exist ok=True)
             # TODO: Your Code Here
             x = tf.cast(img["image"], tf.float32)[None, ...] / 255.0
             t = x
             for layer in model.layers:
                 try:
                     t = layer(t, training = False)
                 except TypeError:
                     t = layer(t)
                 arr = t.numpy()
                 out path = os.path.join(file dir, f"{layer.name}.bin")
                 with open(out_path, "wb") as f:
                     f.write(arr.tobytes())
```

### **Tensorboard**

```
In [18]: tf.config.optimizer.set_jit(False)
    tf.config.run_functions_eagerly(False)

# Setup for profiling
    opts = tf.profiler.experimental.ProfilerOptions(
        host_tracer_level=2, python_tracer_level=0, device_tracer_level=1
)

log_dir = os.path.abspath(os.path.join('log_data'))
log_dir_run = os.path.abspath(os.path.join(log_dir, datetime.datetime.now().pathlib.Path(log_dir_run).mkdir(exist_ok=True, parents=True)

try:
    tf.profiler.experimental.stop()
except:
    test = 2
```

```
finally:
   test = 1
```

```
In [19]: # TODO: Sample Profiling - Inference for a single image:
         import time
         # tf.profiler.experimental.start(log dir run)
         latency = []
         @tf.function(jit compile=False)
         def run infer(x):
             return model(x, training = False)
         _ = run_infer(tf.zeros((1,64,64,3), tf.float32))
         # Perform the inference profiling:
         for i, image in enumerate(sample imgs):
             # Starts Profile logging
             sub dir = os.path.join(log dir run, f'img {i}')
             x = tf.cast(image["image"], tf.float32)
             x = tf.expand dims(x, axis = 0) / 255.0
             tf.profiler.experimental.start(sub dir, options=opts)
             try:
                 tf.profiler.experimental.Trace("single inference", step num = 0, r
                 t0 = time.perf counter()
                 pred = run infer(x)
                 = pred.numpy()
                 t1 = time.perf counter()
                 latency.append((t1-t0) * 1e3)
             finally:
                 tf.profiler.experimental.stop()
                 print(f'Latency for Image {i}: {latency[i]}')
             file writer = tf.summary.create file writer(sub dir)
             with file writer.as default():
                 tf.summary.scalar("latency ms", latency[i], step=0)
             file writer.flush()
```

```
2025-09-08 12:07:49.815893: I external/local_tsl/tsl/profiler/lib/profiler_s ession.cc:104] Profiler session initializing.
2025-09-08 12:07:49.815914: I external/local_tsl/tsl/profiler/lib/profiler_s ession.cc:119] Profiler session started.
2025-09-08 12:07:49.818496: I external/local_tsl/tsl/profiler/lib/profiler_s ession.cc:70] Profiler session collecting data.
2025-09-08 12:07:49.819597: I external/local_tsl/tsl/profiler/lib/profiler_s ession.cc:131] Profiler session tear down.
2025-09-08 12:07:49.822720: I external/local_tsl/tsl/profiler/rpc/client/sav e_profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre_5870/lab1/Lab1/log_data/20250908-120749/img_0/plugins/profile/2025_09_08_12_07_4 9/co2050-19.ece.iastate.edu.xplane.pb
```

```
Latency for Image 1: 6.276489999436308
Latency for Image 2: 7.662865999009227
2025-09-08 12:07:50.057950: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:07:50.057982: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
2025-09-08 12:07:50.064340: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:07:50.064794: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:07:50.067260: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/img 1/plugins/profile/2025 09 08 12 07 5
0/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:07:50.073327: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:07:50.073365: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
2025-09-08 12:07:50.081177: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:07:50.081823: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:07:50.084451: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/img 2/plugins/profile/2025 09 08 12 07 5
0/co2050-19.ece.iastate.edu.xplane.pb
```

Latency for Image 0: 2.2077089997765142

```
In [20]: # Load the TensorBoard notebook extension.
%load_ext tensorboard

# Launch TensorBoard and navigate to the Profile tab to view performance pro
# *** Please note just execute this command once in a session and
# then logs for subsequent runs would be auto detected in tensorboard- url:
print(log_dir_run)
%tensorboard --logdir={log_dir_run} --port=6006

# You can view the tensorboard in the browser url: http://localhost:6006/
# Useful command line to have if tensorboard is misbehaving: kill $(ps -e |
```

/home/buenting/cpre\_5870/lab1/Lab1/log\_data/20250908-120749

```
In [21]: # TODO: Sample Profiling - Online Inference:

# Vary this from 10, 100, 1000 to simulate multiple online inference
loop_index = [10, 100, 1000]

for idx in loop_index:
    sub_dir = os.path.join(log_dir_run, f'loop_{idx}')
    tf.profiler.experimental.start(sub_dir, options=opts)
```

```
for img in ds val.batch(1).take(idx):
         x = tf.cast(img['image'], tf.float32)/255.0
         tf.profiler.experimental.Trace("Multiple Online Inference Profiling"
         pred = run infer(x)
         _ = pred.numpy()
     tf.profiler.experimental.stop()
     print(f'latency test for set {idx} done')
     file writer = tf.summary.create file writer(sub dir)
     with file writer.as default():
         tf.summary.scalar("latency ms", latency[i], step=0)
     file writer.flush()
2025-09-08 12:07:56.196756: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:07:56.196816: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
2025-09-08 12:07:56.288247: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:07:56.290360: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:07:56.302570: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:07:56.305471: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre_5870/
lab1/Lab1/log data/20250908-120749/loop 10/plugins/profile/2025 09 08 12 07
56/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:07:56.313743: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:07:56.313755: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
latency test for set 10 done
2025-09-08 12:07:56.639491: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:07:56.640322: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:07:56.661568: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:07:56.663863: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/loop 100/plugins/profile/2025 09 08 12 07
56/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:07:56.678646: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:07:56.678660: I external/local tsl/tsl/profiler/lib/profiler s
```

latency test for set 100 done

ession.cc:119] Profiler session started.

2025-09-08 12:08:00.356905: W tensorflow/core/framework/local\_rendezvous.cc: 404] Local rendezvous is aborting with status: OUT\_OF\_RANGE: End of sequence 2025-09-08 12:08:00.357790: I external/local\_tsl/tsl/profiler/lib/profiler\_s ession.cc:70] Profiler session collecting data. 2025-09-08 12:08:00.485052: I external/local\_tsl/tsl/profiler/lib/profiler\_s ession.cc:131] Profiler session tear down. 2025-09-08 12:08:00.488109: I external/local\_tsl/tsl/profiler/rpc/client/sav e\_profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre\_5870/lab1/Lab1/log\_data/20250908-120749/loop\_1000/plugins/profile/2025\_09\_08\_12\_08\_00/co2050-19.ece.iastate.edu.xplane.pb

latency test for set 1000 done

```
In [22]: # TODO: Sample Profiling - Batch Inference:
         # We would only perform batch inference for a subset of validation set i.e.
         # using different batch sizes of 20, 40, 100, 200
         # Decides the size of the batch. Try: 20, 40, 100, 200
         batch size = [20, 40, 100, 200]
         for batch in batch size:
             sub dir = os.path.join(log dir run, f'batch {batch}')
             tf.profiler.experimental.start(sub dir, options=opts)
             for imgs in ds val.take(1000).batch(batch):
                 imgs stzd = tf.cast(imgs['image'], tf.float32)/255.0
                 tf.profiler.experimental.Trace("Batch Inference Profiling", step num
                 pred = run infer(imgs stzd)
                 = pred.numpy()
             tf.profiler.experimental.stop()
             print(f'latency test for batch {batch} done')
             file writer = tf.summary.create file writer(sub dir)
             with file writer.as default():
                 tf.summary.scalar("latency ms", latency[i], step=0)
             file writer.flush()
```

```
2025-09-08 12:08:00.630169: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:08:00.630242: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
2025-09-08 12:08:01.245912: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:08:01.247042: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:08:01.276035: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:08:01.278632: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/batch 20/plugins/profile/2025 09 08 12 08
01/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:08:01.300430: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:08:01.300446: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
latency test for batch 20 done
```

```
2025-09-08 12:08:01.782411: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:08:01.783737: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:08:01.810609: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:08:01.813037: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/batch 40/plugins/profile/2025 09 08 12 08
01/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:08:01.834157: I external/local_tsl/tsl/profiler/lib/profiler_s
ession.cc:104] Profiler session initializing.
2025-09-08 12:08:01.834175: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
latency test for batch 40 done
2025-09-08 12:08:02.293801: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:08:02.294883: I external/local_tsl/tsl/profiler/lib/profiler_s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:08:02.319076: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:08:02.321490: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/batch 100/plugins/profile/2025 09 08 12 0
8 02/co2050-19.ece.iastate.edu.xplane.pb
2025-09-08 12:08:02.339561: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:104] Profiler session initializing.
2025-09-08 12:08:02.339574: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:119] Profiler session started.
latency test for batch 100 done
latency test for batch 200 done
2025-09-08 12:08:02.760232: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
2025-09-08 12:08:02.761261: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:70] Profiler session collecting data.
2025-09-08 12:08:02.785159: I external/local tsl/tsl/profiler/lib/profiler s
ession.cc:131] Profiler session tear down.
2025-09-08 12:08:02.787648: I external/local tsl/tsl/profiler/rpc/client/sav
e profile.cc:144] Collecting XSpace to repository: /home/buenting/cpre 5870/
lab1/Lab1/log data/20250908-120749/batch 200/plugins/profile/2025 09 08 12 0
8 02/co2050-19.ece.iastate.edu.xplane.pb
```

### **Training**

```
In [23]: # Setup for model training
    from tensorflow.keras import Model, datasets
    from tensorflow.keras.models import Sequential
    from tensorflow.keras.losses import categorical_crossentropy
    from tensorflow.keras.optimizers import SGD
    from tensorflow.keras.layers import Dense, Flatten, Conv2D, AveragePooling2D
    train_dir = os.path.abspath(os.path.join('train_data', datetime.datetime.now
    pathlib.Path(train_dir).mkdir(exist_ok=True, parents=True)
```

```
# Using early stopping to monitor validation accuracy
callbacks = [
    tf.keras.callbacks.EarlyStopping(
        # Stop training when `val_loss` is no longer improving
        monitor="val_loss",
        # "no longer improving" being defined as "no better than le-2 less"
        min_delta=le-2,
        # "no longer improving" being further defined as "for at least 2 epc
    patience=2,
        verbose=1,
    ),
    tf.keras.callbacks.TensorBoard(log_dir=train_dir, histogram_freq=1)
]
```

```
In [24]: # Basic CNN model
         train model = Sequential()
         # conv1
         train model.add(Conv2D(32, (5, 5), input shape=(64, 64, 3), activation='relu
         train model.add(Conv2D(32, (5,5),activation='relu'))
         train model.add(MaxPooling2D(pool size=(2, 2)))
         train model.add(Conv2D(64, (3,3), activation='relu'))
         train model.add(Conv2D(64, (3,3), activation='relu'))
         train model.add(MaxPooling2D(pool size=(2, 2)))
         train model.add(Conv2D(64, (3,3), activation='relu'))
         train model.add(Conv2D(128, (3,3), activation='relu'))
         train model.add(MaxPooling2D(pool size=(2, 2)))
         train model.add(Flatten())
         # fc1
         train model.add(Dense(256, activation='relu'))
         # fc2
         train model.add(Dense(200, activation='softmax'))
         train model.compile(loss='categorical crossentropy', optimizer='adam', metri
         # TODO: Consider looking at different optimizers and learning rate settings
         train model.summary()
```

```
/home/buenting/cpre_5870/lab1/Lab1/lab1_venv/lib64/python3.9/site-packages/k
eras/src/layers/convolutional/base_conv.py:113: 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 instea
d.
    super().__init__(activity_regularizer=activity_regularizer, **kwargs)
```

Model: "sequential"

Layer (type)	Output Shape	Par
conv2d (Conv2D)	(None, 60, 60, 32)	2
conv2d_1 (Conv2D)	(None, 56, 56, 32)	25
max_pooling2d (MaxPooling2D)	(None, 28, 28, 32)	
conv2d_2 (Conv2D)	(None, 26, 26, 64)	18
conv2d_3 (Conv2D)	(None, 24, 24, 64)	36
<pre>max_pooling2d_1 (MaxPooling2D)</pre>	(None, 12, 12, 64)	
conv2d_4 (Conv2D)	(None, 10, 10, 64)	36
conv2d_5 (Conv2D)	(None, 8, 8, 128)	73
<pre>max_pooling2d_2 (MaxPooling2D)</pre>	(None, 4, 4, 128)	
flatten (Flatten)	(None, 2048)	
dense (Dense)	(None, 256)	524
dense_1 (Dense)	(None, 200)	51

Total params: 770,216 (2.94 MB)

Trainable params: 770,216 (2.94 MB)

Non-trainable params: 0 (0.00 B)

```
In [ ]: # TODO: Attempt to train your own model with different batch sizes
        # TODO: See how long this takes without a GPU on your VDI or 2050 Coover made
        # TODO: THEN log in to your GPU VM, set ENABLE GPU = False in the very first
        # TODO: Make sure you have exported the LD LIBRARY PATH as the lab manual in
        def normalize img(image, label):
            return tf.cast(image, tf.float32) / 255., label
        def to categorical(image, label):
            label = tf.one hot(tf.cast(label, tf.int32), 200)
            return tf.cast(image, tf.float32), tf.cast(label, tf.int64)
        ds re = tiny imagenet builder.as dataset(as supervised=True)
        ds retrain, ds reval = ds re["train"], ds re["validation"]
        ds retrain = ds retrain.cache().shuffle(1024)
        ds reval = ds reval.cache().shuffle(1024)
        ds retrain = ds retrain.map(normalize img, num parallel calls=tf.data.AUTOTU
        ds reval = ds reval.map(normalize img, num parallel calls=tf.data.AUTOTUNE)
        ds retrain = ds retrain.map(to categorical, num parallel calls=tf.data.AUTOT
        ds reval = ds reval.map(to categorical, num parallel calls=tf.data.AUTOTUNE)
        epoch size = 20
```

```
init weights = train model.get weights()
 for batch size in [32, 64, 128]:
     # Setup our batched datasets
     ds retrain batches = ds retrain.batch(batch size)
     ds reval batches = ds reval.batch(batch size)
     train model.fit(
         ds retrain batches,
         epochs = epoch size,
         validation data = ds reval batches,
         callbacks = callbacks
     )
     train model.save(os.path.join(train dir, f'CNN Train Batch Size {batch s
     total = ds reval.cardinality().numpy()
     for images, labels in ds reval.batch(batch size):
         preds = train model.predict(tf.cast(images, tf.float32)/255.0, verbd
         labels ids = np.argmax(labels.numpy(), axis = 1)
         for i, pred in enumerate(preds):
             top 1 = np.argmax(pred)
             if top 1 == labels ids[i]:
                 acc top1 += 1
             top 5 = tf.math.top k(pred, k=5).indices.numpy()
             if labels ids[i] in top 5:
                 acc top5 += 1
     print(f"Top 1 Accuracy: {acc top1/total}\n")
     print(f"Top 5 Accuracy: {acc top5/total}\n")
     train model.set weights(init weights)
Epoch 1/20
3125/3125 •
                       129s 41ms/step - accuracy: 0.0048 - loss: 5.2
992 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 2/20
                            — 144s 46ms/step - accuracy: 0.0046 - loss: 5.2
3125/3125
992 - val accuracy: 0.0050 - val_loss: 5.2984
Epoch 3/20
                         144s 46ms/step - accuracy: 0.0046 - loss: 5.2
3125/3125 -
994 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 3: early stopping
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my model.
keras')` or `keras.saving.save model(model, 'my model.keras')`.
```

```
Top 1 Accuracy: 0.248
       Top 5 Accuracy: 0.519
       Epoch 1/20
          1/1563 -
                                  4:16 164ms/step - accuracy: 0.0000e+00 - los
       s: 5.2995
       2025-09-08 12:15:17.750929: W tensorflow/core/framework/local rendezvous.cc:
       404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
                               136s 87ms/step - accuracy: 0.0048 - loss: 5.3
       001 - val accuracy: 0.0050 - val loss: 5.2983
       Epoch 2/20
       1563/1563 -
                              137s 88ms/step - accuracy: 0.0044 - loss: 5.2
       989 - val accuracy: 0.0050 - val loss: 5.2984
       Epoch 3/20
       137s 88ms/step - accuracy: 0.0047 - loss: 5.2
       991 - val accuracy: 0.0050 - val loss: 5.2984
       Epoch 3: early stopping
       WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
        `keras.saving.save model(model)`. This file format is considered legacy. We
        recommend using instead the native Keras format, e.g. `model.save('my model.
       keras')` or `keras.saving.save model(model, 'my model.keras')`.
       Top 1 Accuracy: 0.253
       Top 5 Accuracy: 0.544
       Epoch 1/20
       2025-09-08 12:22:20.143019: W tensorflow/core/framework/local rendezvous.cc:
       404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
       782/782 — 135s 172ms/step - accuracy: 0.0048 - loss: 5.29
       97 - val accuracy: 0.0050 - val loss: 5.2983
       Epoch 2/20
       782/782 -
                             137s 176ms/step - accuracy: 0.0044 - loss: 5.29
       87 - val accuracy: 0.0050 - val loss: 5.2983
       Epoch 3/20
       782/782 ---
                             139s 177ms/step - accuracy: 0.0041 - loss: 5.29
       88 - val_accuracy: 0.0050 - val_loss: 5.2983
       Epoch 3: early stopping
       WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
        `keras.saving.save_model(model)`. This file format is considered legacy. We
        recommend using instead the native Keras format, e.g. `model.save('my model.
       keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
       Top 1 Accuracy: 0.258
       Top 5 Accuracy: 0.569
       2025-09-08 12:29:19.773799: W tensorflow/core/framework/local rendezvous.cc:
       404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
In [26]: # TODO: Train your model with 3 different numbers of epochs
         batch size = 32
         # Setup your datasets
         ds retrain batches = ds retrain.batch(batch size)
         ds reval batches = ds reval.batch(batch size)
```

```
for epoch size in [3, 10, 100]:
     train model.fit(
         ds retrain batches,
         epochs = epoch size,
         validation_data = ds_reval batches,
         callbacks = callbacks
     )
     # Save the cnn model
     train model.save(os.path.join(train dir, f'CNN Train Epoch Size {epoch s
     #Get top 1 and top 5 percentages
     for images, labels in ds reval.batch(batch size):
         preds = train model.predict((tf.cast(images, tf.float32)/255.0), ver
         labels ids = np.argmax(labels.numpy(), axis = 1)
         for i, pred in enumerate(preds):
             top 1 = np.argmax(pred)
             if top 1 == labels ids[i]:
                acc top1 += 1
             top 5 = tf.math.top k(pred, k=5).indices.numpy()
             if labels ids[i] in top 5:
                 acc top5 += 1
     print(f"Top 1 Accuracy: {acc top1/total}\n")
     print(f"Top 5 Accuracy: {acc top5/total}\n")
     train model.set weights(init weights)
Epoch 1/3
                     142s 45ms/step - accuracy: 0.0044 - loss: 5.3
3125/3125 -
005 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 2/3
            141s 45ms/step - accuracy: 0.0042 - loss: 5.2
3125/3125 -
992 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 3/3
3125/3125 ————
                     142s 45ms/step - accuracy: 0.0044 - loss: 5.2
994 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 3: early stopping
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my model.
keras')` or `keras.saving.save model(model, 'my model.keras')`.
Top 1 Accuracy: 0.263
Top 5 Accuracy: 0.594
Epoch 1/10
                        2:17 44ms/step - accuracy: 0.0191 - loss: 5.3
   3/3125 -
195
2025-09-08 12:36:42.796042: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
```

```
— 142s 45ms/step - accuracy: 0.0049 - loss: 5.3
001 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 2/10
3125/3125 -
                           — 144s 46ms/step - accuracy: 0.0046 - loss: 5.2
993 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 3/10
3125/3125 -
                            - 144s 46ms/step - accuracy: 0.0047 - loss: 5.2
994 - val accuracy: 0.0050 - val loss: 5.2985
Epoch 3: early stopping
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my model.
keras')` or `keras.saving.save model(model, 'my model.keras')`.
Top 1 Accuracy: 0.268
Top 5 Accuracy: 0.619
Epoch 1/100
                          2:01 39ms/step - accuracy: 0.0000e+00 - loss:
   3/3125 -
5.4552
2025-09-08 12:44:10.273935: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
                      142s 45ms/step - accuracy: 0.0046 - loss: 5.3
008 - val accuracy: 0.0050 - val_loss: 5.2984
Epoch 2/100
3125/3125 -
                     142s 45ms/step - accuracy: 0.0046 - loss: 5.2
993 - val accuracy: 0.0050 - val loss: 5.2984
Epoch 3/100
3125/3125 — 143s 46ms/step - accuracy: 0.0044 - loss: 5.2
994 - val accuracy: 0.0050 - val loss: 5.2985
Epoch 3: early stopping
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my model.
keras')` or `keras.saving.save model(model, 'my model.keras')`.
Top 1 Accuracy: 0.273
Top 5 Accuracy: 0.644
2025-09-08 12:51:35.180333: W tensorflow/core/framework/local rendezvous.cc:
404] Local rendezvous is aborting with status: OUT OF RANGE: End of sequence
```

# Above and Beyond

```
In [27]: # Benchmark our dataset to make sure loading our data isn't a bottleneck ...
# (This can be skipped since it can take a bit and is't all that important)
# tfds.benchmark(ds_train.batch(32), batch_size=32, num_iter=2**20)
In [28]: # Explore new models to find a higher-accuracy model. Does the new model required.
```