

```
import tensorflow as tf
import tensorflow_datasets as tfds
import matplotlib.pyplot as plt
import tensorflow_hub as hub
import math,os,datetime
```

```
BATCH_SIZE = 32
```

```
train_data,train_info = tfds.load('cifar10',split='train[10%:90%]',with_info=True)
val_data = tfds.load('cifar10',split='train[0%:10%]')
test_data = tfds.load('cifar10',split='test')
print(train_data)
```

Downloading and preparing dataset cifar10/3.0.2 (download: 162.17 MiB, generated: 132.4

DI Completed....: 100% 1/1 [00:06<00:00, 6.01s/ url]

DI Size....: 100% 162/162 [00:05<00:00, 27.14 MiB/s]

Extraction completed....: 100% 1/1 [00:05<00:00, 5.93s/ file]

Shuffling and writing examples to /root/tensorflow_datasets/cifar10/3.0.2.incomplete3JE
92% 45773/50000 [00:03<00:00, 67406.34 examples/s]

Shuffling and writing examples to /root/tensorflow_datasets/cifar10/3.0.2.incomplete3JE
0% 0/10000 [00:00<?, ? examples/s]

Dataset cifar10 downloaded and prepared to /root/tensorflow_datasets/cifar10/3.0.2. Sub
<PrefetchDataset shapes: {id: (), image: (32, 32, 3), label: ()}, types: {id: tf.string

```
num_train_data = 0
for _ in train_data:
    num_train_data +=1
print(num_train_data)
```

40000

```
num_val_data = 0
for _ in val_data:
    num_val_data+=1
print(num_val_data)
```

5000

```

train_steps_per_epoch = math.ceil(num_train_data/BATCH_SIZE)
val_steps_per_epoch = math.ceil(num_val_data/BATCH_SIZE)

def normalizer(features,input_shape = [224,224,3],augment=True,seed=42):
    input_shape = tf.convert_to_tensor(input_shape)
    image = features['image']
    image = tf.image.convert_image_dtype(image,tf.float32)
    if augment:
        ## Randomly applied horizontal flip:
        image = tf.image.random_flip_left_right(image, seed=seed)

        # Random B/S changes:
        image = tf.image.random_brightness(image, max_delta=0.1, seed=seed)
        image = tf.image.random_saturation(image, lower=0.5, upper=1.5, seed=seed)
        image = tf.clip_by_value(image, 0.0, 1.0) # keeping pixel values in check

        # Random resize and random crop back to expected size:

        random_scale_factor = tf.random.uniform([1], minval=1., maxval=1.4, dtype=tf.float32,
        scaled_height = tf.cast(tf.cast(input_shape[0], tf.float32) * random_scale_factor,
                                tf.int32)
        scaled_width = tf.cast(tf.cast(input_shape[1], tf.float32) * random_scale_factor,
                                tf.int32)
        scaled_shape = tf.squeeze(tf.stack([scaled_height, scaled_width]))
        image = tf.image.resize(image, scaled_shape)
        image = tf.image.random_crop(image, input_shape, seed=seed)
    else:
        image = tf.image.resize(image,input_shape[:2])
    label = features['label']
    features = (image,label)
    return features

train_data = train_data.map(normalizer)
val_data = val_data.map(normalizer)

print(train_data)
print(val_data)

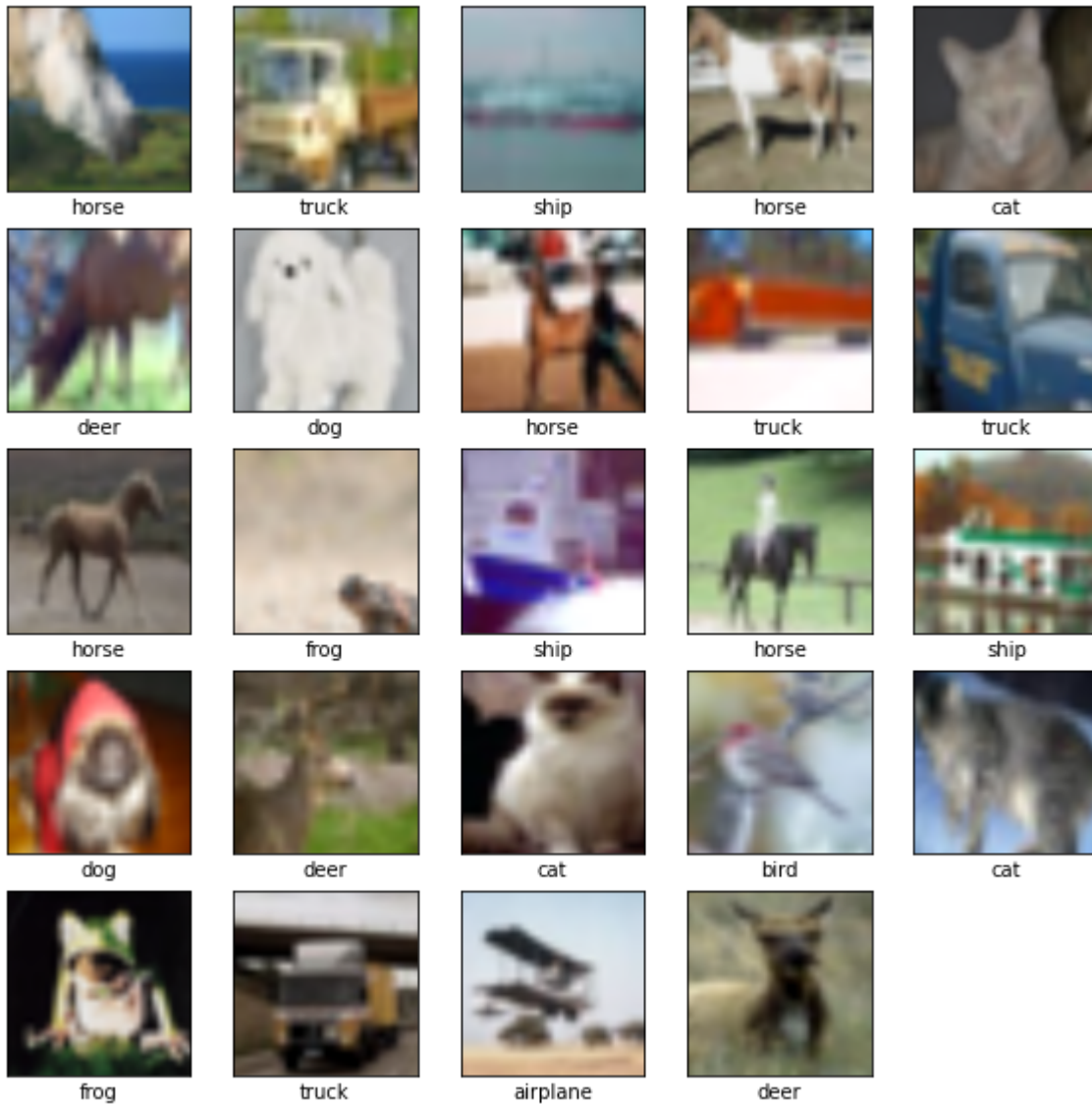
<MapDataset shapes: ((224, 224, 3), ()), types: (tf.float32, tf.int64)>
<MapDataset shapes: ((224, 224, 3), ()), types: (tf.float32, tf.int64)>

class_names = train_info.features["label"].names
print(class_names)

['airplane', 'automobile', 'bird', 'cat', 'deer', 'dog', 'frog', 'horse', 'ship', 'truck']

```

```
plt.figure(figsize=(10,10))
for i , (image,label) in enumerate(train_data.take(24)):
    #image = image.numpy().reshape([28,28,3])
    plt.subplot(5,5,i+1)
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    plt.imshow(image,cmap=plt.cm.binary)
    plt.xlabel(class_names[label])
plt.show()
```



```
train_data = train_data.batch(BATCH_SIZE)
val_data = val_data.batch(BATCH_SIZE)
train_data = train_data.prefetch(1)
val_data = val_data.prefetch(1)
#train_data = train_data.cache()
```

```
print(val_data)
```

```
print(val_data)
print(train_data)
```

```
<PrefetchDataset shapes: ((None, 224, 224, 3), (None,)), types: (tf.float32, tf.int64)>
<PrefetchDataset shapes: ((None, 224, 224, 3), (None,)), types: (tf.float32, tf.int64)>
```

```
Resnet = tf.keras.applications.ResNet50()
```

```
Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/resnet50/102973440/102967424 [=====] - 1s 0us/step
```

```
Resnet.summary()
```

conv5_block1_3_bn (BatchNormali	(None, 7, 7, 2048)	8192	conv5_block1_3_conv
conv5_block1_add (Add)	(None, 7, 7, 2048)	0	conv5_block1_0_bn[0] conv5_block1_3_bn[0]
conv5_block1_out (Activation)	(None, 7, 7, 2048)	0	conv5_block1_add[0]
conv5_block2_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	conv5_block1_out[0]
conv5_block2_1_bn (BatchNormali	(None, 7, 7, 512)	2048	conv5_block2_1_conv
conv5_block2_1_relu (Activation	(None, 7, 7, 512)	0	conv5_block2_1_bn[0]
conv5_block2_2_conv (Conv2D)	(None, 7, 7, 512)	2359808	conv5_block2_1_relu
conv5_block2_2_bn (BatchNormali	(None, 7, 7, 512)	2048	conv5_block2_2_conv
conv5_block2_2_relu (Activation	(None, 7, 7, 512)	0	conv5_block2_2_bn[0]
conv5_block2_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624	conv5_block2_2_relu
conv5_block2_3_bn (BatchNormali	(None, 7, 7, 2048)	8192	conv5_block2_3_conv
conv5_block2_add (Add)	(None, 7, 7, 2048)	0	conv5_block1_out[0] conv5_block2_3_bn[0]
conv5_block2_out (Activation)	(None, 7, 7, 2048)	0	conv5_block2_add[0]
conv5_block3_1_conv (Conv2D)	(None, 7, 7, 512)	1049088	conv5_block2_out[0]
conv5_block3_1_bn (BatchNormali	(None, 7, 7, 512)	2048	conv5_block3_1_conv
conv5_block3_1_relu (Activation	(None, 7, 7, 512)	0	conv5_block3_1_bn[0]
conv5_block3_2_conv (Conv2D)	(None, 7, 7, 512)	2359808	conv5_block3_1_relu
conv5_block3_2_bn (BatchNormali	(None, 7, 7, 512)	2048	conv5_block3_2_conv
conv5_block3_2_relu (Activation	(None, 7, 7, 512)	0	conv5_block3_2_bn[0]

conv5_block3_3_conv (Conv2D)	(None, 7, 7, 2048)	1050624	conv5_block3_2_relu
conv5_block3_3_bn (BatchNormali	(None, 7, 7, 2048)	8192	conv5_block3_3_conv
conv5_block3_add (Add)	(None, 7, 7, 2048)	0	conv5_block2_out[0] conv5_block3_3_bn[0]
conv5_block3_out (Activation)	(None, 7, 7, 2048)	0	conv5_block3_add[0]
avg_pool (GlobalAveragePooling2	(None, 2048)	0	conv5_block3_out[0]
predictions (Dense)	(None, 1000)	2049000	avg_pool[0][0]
=====			
Total params: 25,636,712			
Trainable params: 25,583,592			
Non-trainable params: 53,120			

```

model_dir = './models/Resnet'
logdir = os.path.join("logs", datetime.datetime.now().strftime("%Y%m%d-%H%M%S"))
callbacks = [
    # Callback to interrupt the training if the validation loss/metrics stops improving for s
    tf.keras.callbacks.EarlyStopping(patience=8, monitor='val_acc',
                                     restore_best_weights=True),
    # Callback to log the graph, losses and metrics into TensorBoard:

    tf.keras.callbacks.TensorBoard(model_dir, histogram_freq=1, write_graph=True)

    # Callback to simply log metrics at the end of each epoch (saving space compared to verbose
]
Resnet.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=
               [tf.keras.metrics.SparseCategoricalAccuracy(name='acc'),
                tf.keras.metrics.SparseTopK_categorical_accuracy(k=5, name='top5_acc')])

```

```

history = Resnet.fit(train_data, epochs=20, steps_per_epoch=train_steps_per_epoch, validation_data=

```

```


Epoch 1/20
1250/1250 [=====] - 471s 346ms/step - loss: 1.3580 - acc: 0.55
Epoch 2/20
1250/1250 [=====] - 438s 350ms/step - loss: 0.6499 - acc: 0.77
Epoch 3/20
1250/1250 [=====] - 437s 350ms/step - loss: 0.4627 - acc: 0.83
Epoch 4/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.3329 - acc: 0.88
Epoch 5/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.2329 - acc: 0.91
Epoch 6/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.1767 - acc: 0.93
Epoch 7/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.1301 - acc: 0.95
Epoch 8/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.1022 - acc: 0.96

```

```

Epoch 9/20
1250/1250 [=====] - 435s 348ms/step - loss: 0.0857 - acc: 0.97
Epoch 10/20
1250/1250 [=====] - 436s 348ms/step - loss: 0.0770 - acc: 0.97
Epoch 11/20
1250/1250 [=====] - 436s 349ms/step - loss: 0.0764 - acc: 0.97
Epoch 12/20
1250/1250 [=====] - 435s 348ms/step - loss: 0.0556 - acc: 0.98
Epoch 13/20
1250/1250 [=====] - 435s 348ms/step - loss: 0.0562 - acc: 0.98
Epoch 14/20
1250/1250 [=====] - 435s 348ms/step - loss: 0.0515 - acc: 0.98
Epoch 15/20
1250/1250 [=====] - 434s 347ms/step - loss: 0.0494 - acc: 0.98
Epoch 16/20
1250/1250 [=====] - 434s 347ms/step - loss: 0.0388 - acc: 0.98
Epoch 17/20
1250/1250 [=====] - 434s 347ms/step - loss: 0.0426 - acc: 0.98
Epoch 18/20
1250/1250 [=====] - 434s 348ms/step - loss: 0.0351 - acc: 0.98
Epoch 19/20
1250/1250 [=====] - 434s 347ms/step - loss: 0.0387 - acc: 0.98
Epoch 20/20
1250/1250 [=====] - 434s 348ms/step - loss: 0.0323 - acc: 0.98

```



```

Resnet.save_weights("lenet_weights",overwrite=True)
Resnet.save('My_ResNet')

```

```

INFO:tensorflow:Assets written to: My_ResNet/assets
INFO:tensorflow:Assets written to: My_ResNet/assets

```

```

test_data = tfds.load('cifar10', split='test')
num_tests=0
for _ in test_data:
    num_tests+=1
print(num_tests)

```

```

test_data = test_data.map(normalizer)
test_data = test_data.batch(BATCH_SIZE)
test_data = test_data.prefetch(1)

```

```


Resnet.evaluate(test_data,batch_size=BATCH_SIZE,verbose=1)

```

```

10000
313/313 [=====] - 30s 94ms/step - loss: 0.9160 - acc: 0.8233 -
[0.9159789085388184, 0.8233000040054321, 0.9865999817848206]

```



```

%load_ext tensorboard

```

```

rm -rf ./logs/

```

```
%tensorboard --logdir models/Resnet
```



TensorBoard

SCALARS

GRAPHS

DIS

INACTIVE

- ☐ Show data download links
- ☐ Ignore outliers in chart scaling

Tooltip sorting
method: default

Smoothing



0.6

Horizontal Axis

STEP

RELATIVE

WALL

Runs

Write a regex to filter runs

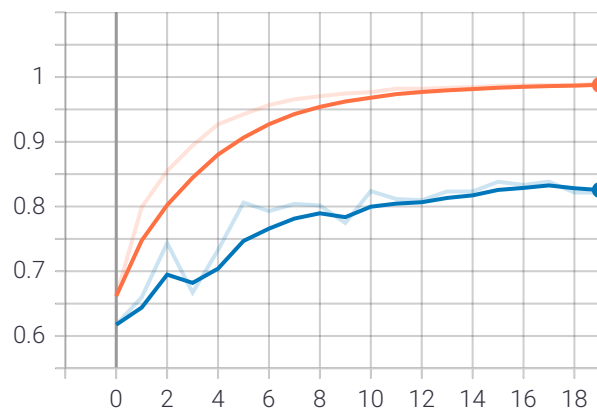
- ☐ ○ train
- ☐ ○ validation

TOGGLE ALL RUNS

models/Resnet

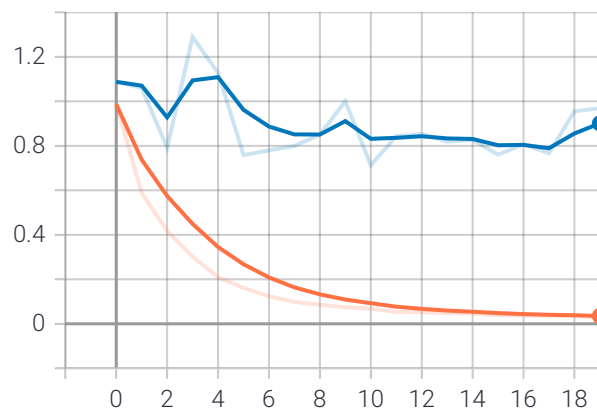
epoch_acc

epoch_acc



epoch_loss

epoch_loss



epoch_top5_acc

✓ 4s completed at 8:17 PM

● ✕