

实验任务

在本实验中，我们将指导大家利用 TensorFlow 完成 MNIST 手写数字识别的任务；同时作为练习，要在每个步骤时，完成 cifar10 的分类任务。

实验预备：

1. 确保实作 1 中已经完成，三个库 TensorFlow, Sonnet, tfmpl 已经安装
2. 阅读材料 <http://yann.lecun.com/exdb/mnist/>，了解 MNIST 数据集
3. 阅读材料 <https://www.cs.toronto.edu/~kriz/cifar.html>，了解 CIFAR 数据集
4. 请在 D 盘下或其它目录下新建文件夹 lab-3

准备训练数据

子任务：

由 MNIST 数据生成特定格式的文件，称为 TensorFlow Record (TFR) 文件，方便训练时使用。TFR 文件是二进制的结构化文件，具有较高的处理效率，可以避免当训练数据过多时，不能加载进内存的困境，又可避免利用一般磁盘文件时的低效率。

步骤：

1. 在 lab-3 文件夹下新建文件 ic_prepare.py
2. 输入以下内容：

```
import os
import numpy as np
import tensorflow as tf

from tensorflow.examples.tutorials.mnist import input_data

def float_feature(value):
    return tf.train.Feature(float_list = tf.train.FloatList(value = value))

def bytes_feature(value):
    return tf.train.Feature(bytes_list = tf.train.BytesList(value = [value]))

def int64_feature(value):
    return tf.train.Feature(int64_list = tf.train.Int64List(value = [value]))

save_path = 'data/mnist'

print("Loading MNIST dataset")
```

```

data = input_data.read_data_sets("data/MNIST/", one_hot=False)

print('beginning prepare MNIST tfrecords for training')
writer = tf.io.TFRecordWriter(os.path.join(save_path, 'mnist-train.tfr'))

num_train_records = 0

for image, label in zip(data.train.images, data.train.labels):
    feature = \
    {
        'image': float_feature(image),
        'label': int64_feature(label)
    }
    example = tf.train.Example(features = tf.train.Features(feature = feature))
    writer.write(example.SerializeToString())

    num_train_records = num_train_records + 1

writer.close()
print('end of tfrecords preparation for training')

print('beginning prepare MNIST tfrecords for testing')
writer = tf.io.TFRecordWriter(os.path.join(save_path, 'mnist-test.tfr'))

num_test_records = 0

for image, label in zip(data.test.images, data.test.labels):
    feature = {
        'image': float_feature(image),
        'label': int64_feature(label)
    }
    example = tf.train.Example(features = tf.train.Features(feature = feature))
    writer.write(example.SerializeToString())

    num_test_records = num_test_records + 1

writer.close()

print('end of tfrecords preparation for testing')

print('#tfrecords for training: {}'.format(num_train_records))
print('#tfrecords for testing: {}'.format(num_test_records))

```



为方便起见，文件 `ic_prepare.py` 附上：

3. 执行文件：D:\lab-3>python ic_prepare.py
4. 查看输出结果，如果程序无误，则会生成如下图所示文件：

OS (D:) > lab-3 > data > MNIST

| Name | Date modified | Type |
|----------------------------|---------------------|----------|
| mnist-test.tfr | 13/10/2021 12:51 PM | TFR File |
| mnist-train.tfr | 13/10/2021 12:51 PM | TFR File |
| t10k-images-idx3-ubyte.gz | 11/10/2021 5:13 PM | GZ File |
| t10k-labels-idx1-ubyte.gz | 11/10/2021 5:13 PM | GZ File |
| train-images-idx3-ubyte.gz | 11/10/2021 5:13 PM | GZ File |
| train-labels-idx1-ubyte.gz | 11/10/2021 5:13 PM | GZ File |

实作：

生成 CIFAR-10 样本的 TensorFlow Record 文件，示例代码已经提供，即 `cifar10.py` 与 `ic_prepare_cifar10.py`，但请读者先自行实验。

注意，实验之前，需要读者先自行下载如下文件放到特定目录：

1. 下载：<https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz>
2. 在当前工作目录下建立 `data\CIFAR-10`，将下载文件放置于此，如：

This PC > OS (D:) > lab-3 > data > CIFAR-10

| Name | Date modified | Type | Size |
|------------------------|--------------------|-------------|------------|
| cifar-10-batches-py | 5/06/2009 6:47 AM | File folder | |
| cifar-10-python.tar.gz | 11/10/2021 7:09 PM | GZ File | 166,503 KB |
| cifar-10-test.tfr | 13/10/2021 5:32 PM | TFR File | 120,528 KB |
| cifar-10-train.tfr | 13/10/2021 5:31 PM | TFR File | 602,637 KB |

观察待分类的图片

子任务：

我们将首先观察待分类的 mnist 图片，以便对我们的任务有一个认识。

步骤:

1. 在 lab-3 文件夹下新建文件 ic_data.py
2. 输入以下内容:

```
import os
import numpy as np
import tensorflow as tf
import sonnet as snt

class Input(snt.AbstractModule):
    def __init__(self, batch_size, image_dims, num_epochs = -1, name = 'input'):
        super(Input, self).__init__(name = name)
        self._batch_size = batch_size
        self._image_dims = image_dims
        self._num_epochs = num_epochs

    def _parse_function(self, example):
        dims = np.prod(self._image_dims)

        features = {
            "image": tf.FixedLenFeature([dims], dtype = tf.float32),
            "label": tf.FixedLenFeature([], dtype = tf.int64)
        }

        example_parsed = tf.parse_single_example(serialized = example, features =
features)
        value = tf.reshape(example_parsed['image'], self._image_dims)

        label = example_parsed['label']

        return value, label

    def _build(self, filename):
        assert os.path.isfile(filename), "invalid file name: {}".format(filename)

        dataset = tf.data.TFRecordDataset([filename])
        dataset = dataset.map(self._parse_function)

        dataset = dataset.batch(self._batch_size)
        dataset = dataset.repeat(self._num_epochs)

        it = dataset.make_one_shot_iterator()
        images, labels = it.get_next()

        return images, labels
```

```

def draw_image(images, rows, cols, tensor_name = "images"):
    import tfmpl

    @tfmpl.figure_tensor
    def draw(images):
        num_figs = len(images)
        fig = tfmpl.create_figures(1, figsize= (12.8, 12.8))[0]

        # pdb.set_trace()
        for i in range(rows):
            for j in range(cols):
                seq = i * cols + j + 1
                if seq > num_figs:
                    fig.tight_layout()
                    return fig

            if num_figs == 1:
                ax = fig.add_subplot(1, 1, 1)
            else:
                ax = fig.add_subplot(rows, cols, seq)

            ax.axis('off')
            ax.imshow(images[seq-1, ...])

        fig.tight_layout()
        return fig

    image_tensor = draw(images)
    image_summary = tf.summary.image(tensor_name, image_tensor)
    sess = tf.get_default_session()
    assert sess != None, "Invalid session"
    image_str = sess.run(image_summary)

    return image_str

if __name__ == '__main__':
    num_images = 64
    image_width = 28
    image_height = 28

    input_ = Input(num_images, [image_height, image_width, 1])

```

```

images, labels = input_('data/mnist/mnist-train.tfr')

if not os.path.exists("output-data"):
    os.makedirs("output-data")

writer = tf.summary.FileWriter("output-data", tf.get_default_graph())

with tf.Session() as sess:
    sess.run([tf.global_variables_initializer(), tf.local_variables_initializer()])

    image_str = draw_image(images, 8, 8)
    writer.add_summary(image_str, global_step = 0)

```



为方便起见，文件 `ic_data.py` 附上：`ic_data.py`

3. 执行文件：D:\lab-3>python ic_data.py
4. 查看输出结果，如果程序无误，则输出如下图所示：
`tensorboard --logdir=D:\\lab-3\\output-data --port=8008`
<http://localhost:8008/#images>



实作：

观察 CIFAR-10 的样本，示例代码，已经提供，但请读者先自行实验。

小结：

可以通过考察真实数据的分布特征，决定所选取模型的复杂度。

建立模型

子任务：

我们用如下配置来建立网络模型：

| 层 ID | 层类型 | Filter 数目 | Filter 大小 | 激活函数 | Padding 方式 |
|------|---------|-----------|-----------|---------|------------|
| 1 | Conv2D | 32 | 5 | ReLU | SAME |
| 2 | Pooling | - | 2 | | - |
| 3 | Conv2D | 64 | 5 | ReLU | SAME |
| 4 | Pooling | - | 2 | | - |
| 5 | FC | - | 256 | ReLU | - |
| 6 | FC | - | #class | Softmax | - |

步骤：

1. 在 lab-3 文件夹下新建文件 ic_model.py
2. 输入以下内容：

```
import os
import numpy as np
import tensorflow as tf
import sonnet as snt
```

```
class Pooling(snt.AbstractModule):
```

```
    def __init__(self, pool = None, k = 2, padding = 'SAME', name = "pooling"):
        super(Pooling, self).__init__(name = name)
        self._pool = pool
        self._k = k
        self._padding = padding
```

```
    def _build(self, x):
```

```
        if self._pool == 'max':
            return tf.nn.max_pool2d(x, self._k, self._k, self._padding)
        elif pool == 'avg':
            return tf.nn.avg_pool2d(x, self._k, self._k, self._padding)
        else:
            return lambda x: x
```

```
class Model(snt.AbstractModule):
```

```
    def __init__(self, num_classes, filter_size = 5, name = "model"):
        super(Model, self).__init__(name = name)
```

```
        with self._enter_variable_scope():
```

```
            self._conv1 = snt.Conv2D(32, filter_size, name = "first_conv_layer")
```

```
self._pool1 = Pooling('max', name = "first_max_pool_layer")
```

```
self._conv2 = snt.Conv2D(64, filter_size, name = "second_conv_layer")
```

```
self._pool2 = Pooling('max', name = "second_pool_layer")
```

```
self._lin = snt.Linear(256, name = "fully_conn_layer")
```

```
self._output = snt.Linear(num_classes, name = "output_layer")
```

```
def _build(self, x):
```

```
    y = tf.nn.relu(self._conv1(x))
```

```
    y = self._pool1(y)
```

```
    y = tf.nn.relu(self._conv2(y))
```

```
    y = self._pool2(y)
```

```
    y = snt.BatchFlatten()(y)
```

```
    y = tf.nn.relu(self._lin(y))
```

```
    return self._output(y)
```

如果要增加新的卷积层与池化层，请在这里定义，如：

```
self._conv3 = snt.Conv2D(128, filter_size,  
name = "third_conv_layer")
```

```
self._pool3 = Pooling('max', name  
= "third_pool_layer")
```

如果定义了新的层，请在这里建构在一起，如 `y = tf.nn.relu(self._conv3(y))` 等等

```
def test():
```

```
    x = tf.random_normal([32, 28, 28, 1], name="x")
```

```
    model = Model(10)
```

```
    y = model(x)
```

```
    if not os.path.exists("output-model"):
```

```
        os.makedirs("output-model")
```

```
    writer = tf.summary.FileWriter("output-model")
```

```
    with tf.Session() as sess:
```

```
        writer.add_graph(sess.graph)
```

```
        sess.run([tf.global_variables_initializer(),
```

```
                  tf.local_variables_initializer()])
```

```
        sess.run(y)
```

```
    writer.close()
```



```
if __name__ == "__main__":  
    test()
```



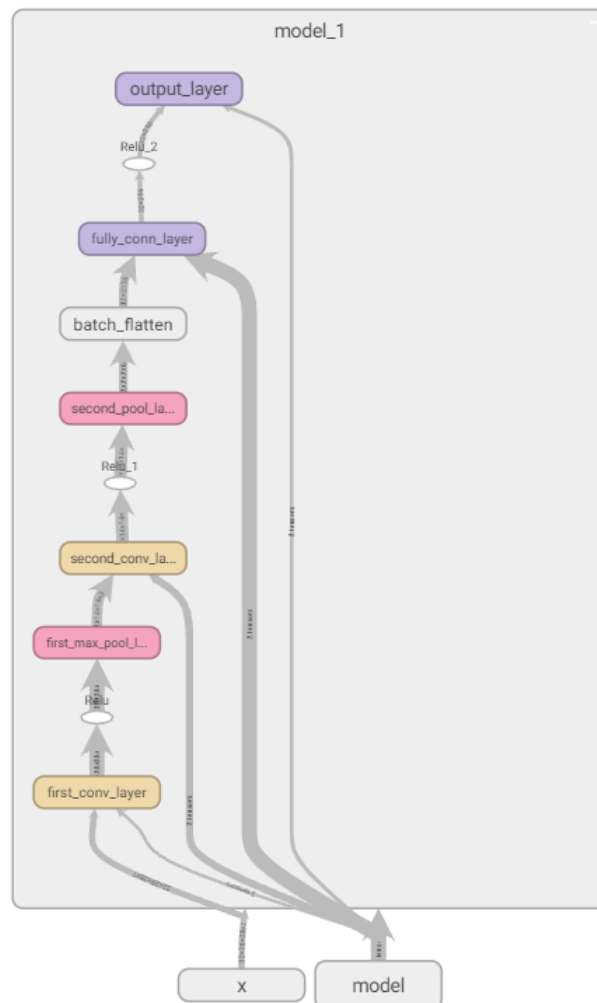
为方便起见，文件 ic_model.py 附后： ic_model.py

3. 查看输出结果，如果程序无误，则输出如下图所示：

D:\lab-3>python ic_model.py

tensorboard --logdir=D:\\lab-3\\output-model --port=8008

[http://localhost:8008/#graphs&run=.](http://localhost:8008/#graphs&run=)



实作：

请以如下配置构建网络，通过 TensorBoard 观察所建立模型是否正确。

| 层 ID | 层类型 | Filter 数目 | Filter 大小 | 激活函数 | Padding 方式 |
|------|---------|-----------|-----------|---------|------------|
| 1 | Conv2D | 32 | 3 | ReLU | SAME |
| 2 | Pooling | - | 2 | | - |
| 3 | Conv2D | 64 | 3 | ReLU | SAME |
| 4 | Pooling | - | 2 | | - |
| 5 | Conv2D | 128 | 3 | ReLU | SAME |
| 6 | Pooling | - | 2 | | - |
| 7 | FC | - | 256 | ReLU | - |
| 8 | FC | - | #class | Softmax | - |

训练模型

子任务：

我们用 mini-batch 方法，训练新建模型。

步骤：

1. 在 lab-3 文件夹下新建文件 ic_train.py
2. 输入以下内容

```
import os
import numpy as np
import tensorflow as tf
```

```
from ic_data import Input
from ic_model import Model
```

```
batch_size = 32
```

```
image_width = 28
image_height = 28
num_classes = 10
```

如果用 cifar-10，则图片是
32x32 的，这里也要改

```
# 执行自动微分算法（或反向传播算法）的优化器参数
```

```
learning_rate = 0.001
lr_decay_steps = 100
lr_decay_factor = 0.9
```

如果观察到训练不收敛，可能学习率设
置过大（或过小），请调整学习率

```
# 训练时迭代次数
```

```
iterations = 1000
```

```
if not os.path.exists("output-train"):
    os.makedirs("output-train")
```

cifar-10 是彩色图片，RGB 共 3 个通道，这里 1 要改成 3

```
# 训练好的模型的保存路径
```

```
checkpoint_path = os.path.join("output-train", "model.ckpt")
```

```
# 以数据集方式加载样本，方便进行 mini-batch 训练
```

```
input_ = Input(batch_size, [image_height, image_width, 1])
```

```
images, labels = input_({"data/mnist/mnist-train.tfr"})
```

```
# 由输入得到模型的输出
```

```
net = Model(num_classes)
```

```
logits = net(images)
```

如果训练 CIFAR-10，请更改为 CIFAR-10 的 tfr 文件所在的正确的路径

```
# 计算损失函数
```

```
labels = tf.one_hot(labels, num_classes, axis = -1)
```

```
with tf.control_dependencies([tf.assert_equal(tf.rank(labels), tf.rank(logits))]):
```

```
    loss = tf.nn.softmax_cross_entropy_with_logits_v2(labels = labels, logits = logits)
```

```
loss = tf.reduce_mean(loss, name = 'loss')
```

```
loss_summary = tf.summary.scalar('loss', loss)
```

```
# 设置学习率
```

```
global_step = tf.train.get_or_create_global_step()
```

```
lr = tf.train.exponential_decay(learning_rate, global_step,
```

```
    lr_decay_steps, lr_decay_factor, staircase = True)
```

```
lr_summary = tf.summary.scalar('lr', lr)
```

```
# 创建优化器
```

```
opt = tf.train.AdamOptimizer(lr)
```

```
# 进行优化
```

```
train_op = opt.minimize(loss, global_step = global_step, var_list =
```

```
tf.trainable_variables())
```

```
# 合并所有 summary 信息
```

```
all_summaries = tf.summary.merge_all()
```

```
writer = tf.summary.FileWriter('output-train', tf.get_default_graph())
```

```

# 保存训练好的模型
saver = tf.train.Saver(tf.trainable_variables())

# 创建 Session, 进行训练
with tf.Session() as sess:
    sess.run([tf.global_variables_initializer(), tf.local_variables_initializer()])

    for i in range(iterations):
        loss_val, _, summ_str = sess.run([loss, train_op, all_summaries])
        print("{}-th iteration with loss {}".format(i, loss_val))
        writer.add_summary(summ_str, global_step = i)

# 训练完成, 保存模型
print('Saving model.')
saver.save(sess, checkpoint_path)
print('Training complete')

writer.close()

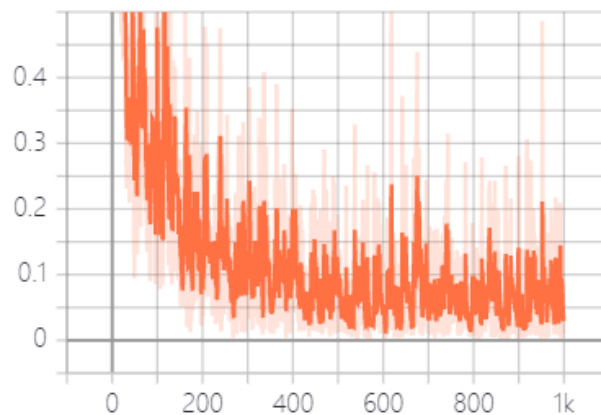
```



为方便起见, 文件 `ic_train.py` 附上: `ic_train.py`

- 查看输出结果, 如果程序无误, 则输出如下图所示:
`D:\lab-3>python ic_train.py`
`tensorboard --logdir=D:\\lab-3\\output-train --port=8008`
<http://localhost:8008/#scalars>

loss_1



实作：

依据上节中所建网络，训练识别 CIFAR-10 图片所用的网络模型。

测试模型

子任务：

测试我们用 mini-batch 方法所训练模型的表现。

步骤：

1. 在 lab-3 文件夹下新建文件 ic_test.py
2. 输入以下内容

```
import os
import numpy as np
import tensorflow as tf
import tfmpl

from ic_data import Input
from ic_model import Model

from tensorflow.contrib.tensorboard.plugins import projector
```

```
image_width = 28
image_height = 28
num_classes = 10
```

如果用 cifar-10，则图片是
32x32 的，这里也要改

```
LOG_DIR = "output-test"
```

```
if not os.path.exists(LOG_DIR):
    os.makedirs(LOG_DIR)
```

```
if not os.path.exists("output-train"):
    raise ValueError("Non-existing output-train folder")
```

```
# 训练好的模型的保存路径
checkpoint_dir = "output-train"
```

```
graph_test = tf.Graph()
with graph_test.as_default():
    # 加载测试数据
    input_ = Input(1, [image_height, image_width, 1], 1)
```

cifar-10 是彩色图片，RGB 共 3 个通
道，这里 1 要改成 3

```
image, label = input_("data/mnist/mnist-test.tfr")
```

```
# 由输入得到模型的输出  
net = Model(num_classes)  
logit = net(image)
```

如果训练 CIFAR-10，请更改为 CIFAR-10
的 tfr 文件所在的正确的路径

```
logit = tf.squeeze(logit, axis = 0)  
id = tf.argmax(logit, axis = -1)
```

```
restorer = tf.train.Saver()
```

```
embeddings = []  
metadata = os.path.join(LOG_DIR, 'metadata.tsv')  
metadata_file = open(metadata, 'w')  
count = 0
```

```
with tf.Session(graph = graph_test) as sess:  
    sess.run([tf.global_variables_initializer(), tf.local_variables_initializer()])
```

```
ckpt = tf.train.get_checkpoint_state(checkpoint_dir)  
if ckpt and ckpt.model_checkpoint_path:  
    # Restores from checkpoint.  
    restorer.restore(sess, ckpt.model_checkpoint_path)
```

```
print('Successfully loaded model from %s.' % ckpt.model_checkpoint_path)
```

```
else:  
    print('No checkpoint file found')  
    exit
```

```
while True:  
    try:  
        logit_val, id_val = sess.run([logit, id])  
        embeddings.append(logit_val)  
        metadata_file.write('%d\n' % id_val)  
        # sess.run(metric_update)  
        count = count + 1  
    except tf.errors.OutOfRangeError:  
        break
```

```
metadata_file.flush()  
metadata_file.close()
```

```

graph_visualize = tf.Graph()
with graph_visualize.as_default():
    embedding_var = tf.Variable(np.zeros([count, num_classes]), dtype = tf.float32,
name="embedding_var")
    embedding_op = embedding_var.assign(tf.convert_to_tensor(np.array(embeddings),
dtype = tf.float32))

    saver = tf.train.Saver([embedding_var])

    writer = tf.summary.FileWriter(LOG_DIR, graph_visualize)
    config = projector.ProjectorConfig()

    embedding = config.embeddings.add()
    embedding.tensor_name = embedding_var.name
    embedding.sprite.image_path = 'mnist_10k_sprite.png' # Path relative to writer's log
directory
    embedding.sprite.single_image_dim.extend([28, 28])
    embedding.metadata_path = 'metadata.tsv' # Path relative to writer's log
directory

    projector.visualize_embeddings(writer, config)

with tf.Session(graph = graph_visualize) as sess:
    sess.run([tf.global_variables_initializer(), tf.local_variables_initializer()])
    sess.run(embedding_op)

    saver.save(sess, os.path.join(LOG_DIR, 'model.ckpt'), global_step = 1)

```



为方便起见，文件 ic_test.py 附后： ic_test.py

3. 执行文件：D:\lab-3>python ic_test.py
4. 将如下 sprite 图片放入目录 output-test 中：

5. 查看输出结果，如果程序无误，则输出如下图所示：

```
tensorboard --logdir=D:\\lab-3\\output-test --port=8008
```

```
tensorboard --logdir=D:\\lab-3\\output-test --port=8008
```


[http://localhost:8008/#projector&run=.](http://localhost:8008/#projector&run=)



实作:

依据上节训练结果，测试 CIFAR-10 的模型的表现。

注意，要将如下图片放入 output-test 文件夹中：

