# 深度学习 Lab6-recurrent neural network

兰韵诗

本次Lab有作业,请在4月7号结束之前提交!!!

## Lab5参考答案

```
class Conv2D(nn.Module):
    def __init__(self, out_channels, in_channels, kernel_size):
        super(Conv2D, self).__init__()
        self.weight = np.random.randn(out_channels, in_channels, kernel_size[0], kernel_size[1])
        self.bias = np.random.randn(out channels)
    def forward_np(self, X):
        X --> (B, I, H, W) where B = batch size, I = in_channel, H = height of feature map, W = width of feature map
        你需要利用以上初始化的参数weight和bias实现一个卷积层的前向传播
        Y should have size (B, 0, H-h+1, W-w+1)
        Y = corr2d(X, self.weight) + np.reshape(self.bias, (1, -1, 1, 1))
        return Y
def corr2d(X, K):
   X --> (B, I, H, W) where B = batch size, I = in_channel, H = height of feature map, W = width of feature map
   K \longrightarrow (0, I, h, w) where 0 = out\_channel, I = in\_channel, h = height of kernel, w = width of kernel
   你需要实现一个Stride为1, Padding为0的窄卷积操作
   Y should have size (B, 0, H-h+1, W-w+1)
   B, I, H, W = X.shape
   o, _, h, w = K.shape
   Y = np.zeros((B, o, H - h + 1, W - w + 1))
   for i in range(Y.shape[-2]):
       for j in range(Y.shape[-1]):
           Y[:, :, i, j] = np.sum(np.sum(np.sum(np.repeat(np.expand_dims(X[:, :, i: i + h, j: j + w], axis = 2), repeats = 0, axis = 2)
               * np.transpose(np.reshape(K, (1, o, I, h, w)), (0, 2, 1, 3, 4)), -1), -1), 1)
    return Y
```

### Lab5参考答案

```
class MaxPool2D(nn.Module):
   def __init__(self, pool_size):
       super(MaxPool2D, self). init ()
       self.pool_size = pool_size
   def forward_np(self, X):
       X \longrightarrow (B, I, H, W) where B = batch size, I = in_channel, H = height of feature map, W = width of feature map
       K --> (h, w) where h = height of kernel, w = width of kernel
       你需要利用以上pool_size实现一个汇聚层的前向传播,汇聚层的子区域间无覆盖
       Y should have size (B, I, H/h, W/w)
       B, I, H, W = X.shape
       p_h, p_w = self.pool_size
       Y = np.zeros((B, I, int(H/p_h), int(W/p_w)))
       for i in range(Y.shape[-2]):
           for j in range(Y.shape[-1]):
               Y[:, :, i, j] = np.amax(np.amax(X[:, :, i*p_h: (i+1)*p_h, j*p_w: (j+1)*p_w], -1), -1)
       return Y
```

### Lab5参考答案

#### Lab6

- 1.熟悉文本生成任务的流程
- 2.补全rnn\_hard\_version.py 文件中的基于GRU的歌词预测模型

#### Recurrent Neural Network

- 根据提示,补全**基于GRU的歌词预测模型**代码
  - 利用设定好的输入完成GRU的前向传播和歌词预测模型主体
  - ·正确定义和初始化GRU中的参数
  - 不能调用其他工具包,不能调用pytorch内置的GRU模块,只能在 "to do" 下面书写代码
  - 提交之后,测试集上的准确率应该在一个正确的范围内可多次提交。即 使对自己的代码没有自信也一定要提交,我们会酌情给过程分
- TO DO: 完成《Recurrent Neural Network》项目。补全 rnn\_hard\_version.py文件使exercise\_rnn.py文件中的 train\_with\_RNN\_hard()可以顺利执行。

#### Evaluation脚本

```
def compute_acc(pred_file):
   with open('data/jaychou_test_y.txt') as f:
       gold = f.readlines()
   gold = [sent.strip() for sent in gold]
   with open(pred_file) as f:
       pred = f.readlines()
   pred = [sent.strip() for sent in pred]
   correct_case = [i for i, _ in enumerate(gold) if gold[i] == pred[i]]
   acc = len(correct_case)*1./len(gold)
   print('the predicted accuracy is %s' %acc)
  __name__ == '__main__':
   pred_file = 'data/predict.txt'
   compute_acc(pred_file)
```

Note:为了测试方便,这里我们使用准确率作为我们的生成评估标准。实际生成任务一般采用BLEU,Rouge等