

人工智能导论实验3

- 课程名称：人工智能导论实验
- 实验项目名称：人工智能导论实验作业3-基于神经网络的分类实验
- 学生姓名：徐彬涵
- 专业班级：软件工程2003
- 学号：32001272
- 实验成绩：
- 指导老师：李卓蓉
- 日期：2022/05/25

数据准备（使用CIFAR10）

```
1  gt_coord: (1.25, 3.04, 0.80)
2  dists: ( 1.7261, 2.8658, 6.0083, 5.0984, 4.0149)
3
4  fp_level_poll: (-79.8, -94.7, -116.4, -91.2, -93.1)
5
6  rx_level_poll: (-77.7,-83.7,-97.8,-87.9,-84.6)
7
8  fp_level_final: (-80.9,-97.0,-114.5,-90.3,-93.3)
9
10 rx_level_final: (-77.8,-83.2,-97.4,-87.6,-84.3)
11
12 original coord: (149.0868, 323.2389, -2.1992),  filtered coord: (143.7123, 318.8991, -19.5418)
13
14 unselected base:(2)
15
16 dists: ( 1.7495, 2.6782, 6.1068, 5.0656, 4.0525)
17
18 fp_level_poll: (-80.7, -97.3, -116.1, -91.5, -93.5)
19
20 rx_level_poll: (-77.9,-82.8,-97.5,-88.0,-84.5)
21
22 fp_level_final: (-81.2,-97.0,-117.2,-92.9,-93.5)
23
24 rx_level_final: (-77.4,-82.6,-97.5,-88.3,-84.2)
25
26 original coord: (144.5497, 323.3594, -9.2688),  filtered coord: (143.5451, 320.1567, -16.8280)
27
28 unselected base:(2)
```

该类型文件共16个，每组包含50个以上的数据，每条数据包含dists、fp_level_poll.....等数据

```
dataset.py × 0-1.txt × bases.txt × codecs.py × _bootlocale.py × posixpath.py × typing.py ×
bases[0].x = (1.25-0.39); bases[0].y = (3.04-1.5); bases[0].z = 0.6;
bases[1].x = -(1.25-0.43); bases[1].y = (3.04-1.40); bases[1].z = 0.6;
bases[2].x = -(1.25-0.43); bases[2].y = -(3.04-1.28); bases[2].z = 0.6;
bases[3].x = (1.25-0.41); bases[3].y = -(3.04-1.27); bases[3].z = 0.6;
bases[4].x = 0.0; bases[4].y = -0.31; bases[4].z = 0.6;
```

标准参数，用于计算标准距离

```
1 class Measure:
2     def __init__(self, gt_dist, mea_dist, fp_level_poll, rx_level_poll, fp_level_final,
    rx_level_final, base_id, in_out):
3         self.gt_dist = gt_dist
4         self.mea_dist = mea_dist
5         self.fp_level_poll = fp_level_poll
6         self.rx_level_poll = rx_level_poll
7         self.fp_level_final = fp_level_final
8         self.rx_level_final = rx_level_final
9         self.base_id = base_id
10        self.in_out = in_out
11
```

用于存储数据的类

```
1 import torch
2
3 import os
4 import numpy as np
5 from measure import Measure
6
7
8 def calc_dist(a, b):
9     return np.linalg.norm(a - b)
10
11
12 def calc_dists2bases(a, bases):
13     dists = []
14     for b in bases:
15         dists.append(calc_dist(a, b))
16     dists = np.asarray(dists)
17     return dists
18
19
20 class UWBDataset(torch.utils.data.Dataset):
```

```

21 def __init__(self, file_path):
22     self.base_num = 5
23     base_coords = np.zeros((self.base_num, 3))
24     with open(os.path.join(file_path, 'bases.txt'), 'r') as f:
25         lines = f.readlines()
26         for line in lines:
27             lhs = line.split(';')[0].split('=')[0]
28             base_i = int(lhs.split('[')[1].split(']')[0])
29             # print(base_num)
30             sub_strs = line.split(';')
31             val = np.asfarray([eval(sub_str.split('=')[1]) for sub_str in sub_strs if "=" in
sub_str])
32             base_coords[base_i] = val
33             file_list = os.listdir(os.path.join(file_path, 'measures'))
34             self.measures = []
35             self.len = 0
36
37         for file_name in file_list:
38             with open(os.path.join(file_path, 'measures', file_name), 'r') as f:
39                 lines = f.readlines()
40                 started = False
41                 cnt = 0
42                 if file_name[0:2] == 'in':
43                     in_out = 'in'
44                 else:
45                     in_out = 'out'
46                     # continue
47                 # print(in_out)
48                 for line in lines:
49                     try:
50                         first_word = line.split(':')[0]
51                         first_val_strs = line.split('(')[1].split(')')[0].split(',')
52                         first_val_np = np.asfarray([float(str_i) for str_i in first_val_strs])
53                         if first_word == 'gt_coord':
54                             gt_coord = first_val_np
55                             gt_dists = calc_dists2bases(gt_coord, base_coords)
56                         elif first_word == 'dists':
57                             if not started:
58                                 started = True
59                                 mea_dists = first_val_np
60                         elif first_word == 'rx_level_poll':
61                             # print(first_word)

```

```

62         rx_level_polls = first_val_np
63     elif first_word == 'fp_level_poll':
64         fp_level_polls = first_val_np
65     elif first_word == 'fp_level_final':
66         fp_level_finals = first_val_np
67     elif first_word == 'rx_level_final':
68         rx_level_finals = first_val_np
69     if started:
70         tmp_measure = []
71         for i in range(self.base_num):
72             tmp_measure.append(
73                 Measure(gt_dists[i], mea_dists[i], fp_level_polls[i],
74                       rx_level_polls[i],
75                       fp_level_finals[i], rx_level_finals[i], i, in_out))
76             self.measures.append(tmp_measure)
77             cnt += 1
78             if cnt >= 50:
79                 break
80         except Exception as e:
81             pass
82
83     def __len__(self):
84         return len(self.measures)
85
86     def __getitem__(self, idx):
87         x_list = []
88         y_list = []
89         for i in range(self.base_num):
90             x_list.append(self.measures[idx][i].mea_dist)
91             x_list.append(self.measures[idx][i].fp_level_poll)
92             y_list.append(self.measures[idx][i].gt_dist)
93
94         x_array = np.asfarray(x_list, dtype=float)
95         y_array = np.asfarray(y_list, dtype=float)
96         x_tensor = torch.from_numpy(np.float32(x_array))
97         y_tensor = torch.from_numpy(np.float32(y_array))
98
99         return x_tensor, y_tensor

```

模型

```

1  from torch import nn
2  from torch.nn import Sequential, Linear, ReLU, Dropout
3
4
5  class DisFit(nn.Module):
6      def __init__(self):
7          super(DisFit, self).__init__()
8          self.model = Sequential(
9              Linear(10, 24),
10             ReLU(True),
11             Dropout(),
12             Linear(24, 24),
13             ReLU(True),
14             Dropout(),
15             Linear(24, 5)
16         )
17
18     def forward(self, x):
19         x = self.model(x)
20         return x

```

使用了3层网络：

- 线性层 10 -> 24 + ReLU 激活函数 + 正则化层
- 线性层 24 -> 24 + ReLU 激活函数 + 正则化层
- 线性层 24 -> 5

训练

```

1  from shutil import copyfile
2  from dataset import UWBDDataSet
3  from measure import Measure
4  from model import DisFit
5  import torch
6  import argparse
7  from torch.utils.data import DataLoader
8  import os
9  import numpy as np
10 from timeit import default_timer as timer
11 from evaluator import get_accuracy
12 import math
13

```

```
14 parser = argparse.ArgumentParser(description='UWB')
15
16 parser.add_argument('--cuda', type=int, default=0, help='cuda number
17 (default: 1)')
18
19 parser.add_argument('--output_dir', type=str,
20 default='./results/model_weights/main/', help='output_dir')
21
22 parser.add_argument('--data_dir', type=str, default='./data/2022年4月20日静态数
23 据', help='output_dir')
24
25 parser.add_argument('--epochs', type=int, default=200, help='number of
26 epochs (default: 5)')
27
28 parser.add_argument('--batch_size', type=int, default=5, help='batch size for
29 training (default: 32)')
30
31 parser.add_argument('--lr', type=float, default=0.0001, help='learning rate
32 (default: 5e-3)')
33
34
35
36
37 params = parser.parse_args()
38 print(params)
39
40 torch.cuda.set_device(params.cuda)
41
42
43 dataset = UWBDataset(params.data_dir)
44 length = len(dataset)
45 train_size, validate_size = int(0.8*length), int(0.2*length)
46 train_dataset, test_dataset = torch.utils.data.random_split(dataset, [train_size,
47 validate_size])
48
49 train_data_loader = DataLoader(train_dataset, batch_size=params.batch_size,
50 shuffle=True,)
51
52 test_data_loader = DataLoader(test_dataset, batch_size=1)
53
54
55 model = DisFit()
56 model = model.cuda()
57
58
59 loss_fn = torch.nn.MSELoss()
60 loss_fn = loss_fn.cuda()
61
62 optimizer = torch.optim.Adam(model.parameters(), lr=params.lr)
63
64 loss_best = math.inf
65
66
67
68 for epoch in range(params.epochs):
69     start_time = timer()
```

```

48     losses = []
49     model.train()
50     for x, y in train_data_loader:
51         x = x.cuda()
52         y = y.cuda()
53         predict = model(x)
54
55         loss_val = loss_fn(predict, y)
56         loss_val.backward()
57         losses.append(loss_val.data.cpu().numpy())
58         optimizer.step()
59
60     with torch.no_grad():
61         loss = get_accuracy(model, test_data_loader)
62         print(
63             "Epoch {} : 平均训练集Loss: {:.5f}, 验证集Loss: {:.5f}, Time elapsed {:.2f}
s".format(
64             epoch + 1,
65             np.mean(losses),
66             loss,
67             timer() - start_time
68         )
69     )
70     if loss < loss_best:
71         stopping_step = 0
72         print("loss reduced....saving weights !!")
73         best_loss_epoch = epoch + 1
74         loss_best = loss
75         output_dir = params.output_dir
76         os.makedirs(output_dir, exist_ok=True)
77         model_path = output_dir + "epoch{}_loss_{:.5f}.pth".format(epoch + 1, loss)
78         torch.save(model.state_dict(), model_path)
79         print("model saved in " + model_path)
80
81     if epoch + 1 == params.epochs:
82         best_model_path = output_dir +
83         "epoch{}_loss_{:.5f}.pth".format(best_loss_epoch, loss_best)
84         copyfile(best_model_path, output_dir + "final.pth")
85         print("the model " + model_path + " is saved in " + model_path)

```

- 使用 CUDA 训练
- 训练 epochs = 200

- 训练 batch size = 5
- 训练学习率 = 0.0001
- 训练损失函数: MSELoss
- 训练优化器: Adam

训练集、测试集划分

```
1 dataset = UWBDataset(params.data_dir)
2 length = len(dataset)
3 train_size, validate_size = int(0.8*length), int(0.2*length)
4 train_dataset, test_dataset = torch.utils.data.random_split(dataset, [train_size,
    validate_size])
5 train_data_loader = DataLoader(train_dataset, batch_size=params.batch_size,
    shuffle=True,)
6 test_data_loader = DataLoader(test_dataset, batch_size=1)
```

获取16个文件前50条数据，并随机将80%作为训练集，20%作为测试集

结果输出

以测试集的loss为批评标准，找到最小的loss，将局部最优解保存，并在最后保存已找到的最优解