model-sageconv

March 18, 2024

[1]: !pip install torch-geometric

Collecting torch-geometric Downloading torch_geometric-2.5.1-py3-none-any.whl (1.1 MB) 1.1/1.1 MB 7.6 MB/s eta 0:00:00 Requirement already satisfied: tqdm in /usr/local/lib/python3.10/distpackages (from torch-geometric) (4.66.2) Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from torch-geometric) (1.25.2) Requirement already satisfied: scipy in /usr/local/lib/python3.10/dist-packages (from torch-geometric) (1.11.4) Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from torch-geometric) (2023.6.0) Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages (from torch-geometric) (3.1.3) Requirement already satisfied: aiohttp in /usr/local/lib/python3.10/distpackages (from torch-geometric) (3.9.3) Requirement already satisfied: requests in /usr/local/lib/python3.10/distpackages (from torch-geometric) (2.31.0) Requirement already satisfied: pyparsing in /usr/local/lib/python3.10/distpackages (from torch-geometric) (3.1.2) Requirement already satisfied: scikit-learn in /usr/local/lib/python3.10/distpackages (from torch-geometric) (1.2.2) Requirement already satisfied: psutil>=5.8.0 in /usr/local/lib/python3.10/distpackages (from torch-geometric) (5.9.5) Requirement already satisfied: aiosignal>=1.1.2 in /usr/local/lib/python3.10/dist-packages (from aiohttp->torch-geometric) (1.3.1) Requirement already satisfied: attrs>=17.3.0 in /usr/local/lib/python3.10/distpackages (from aiohttp->torch-geometric) (23.2.0) Requirement already satisfied: frozenlist>=1.1.1 in /usr/local/lib/python3.10/dist-packages (from aiohttp->torch-geometric) (1.4.1) Requirement already satisfied: multidict<7.0,>=4.5 in /usr/local/lib/python3.10/dist-packages (from aiohttp->torch-geometric) (6.0.5) Requirement already satisfied: yarl<2.0,>=1.0 in /usr/local/lib/python3.10/distpackages (from aiohttp->torch-geometric) (1.9.4) Requirement already satisfied: async-timeout<5.0,>=4.0 in /usr/local/lib/python3.10/dist-packages (from aiohttp->torch-geometric) (4.0.3)

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Requirement already satisfied: MarkupSafe>=2.0 in
    /usr/local/lib/python3.10/dist-packages (from jinja2->torch-geometric) (2.1.5)
    Requirement already satisfied: charset-normalizer<4,>=2 in
    /usr/local/lib/python3.10/dist-packages (from requests->torch-geometric) (3.3.2)
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-
    packages (from requests->torch-geometric) (3.6)
    Requirement already satisfied: urllib3<3,>=1.21.1 in
    /usr/local/lib/python3.10/dist-packages (from requests->torch-geometric) (2.0.7)
    Requirement already satisfied: certifi>=2017.4.17 in
    /usr/local/lib/python3.10/dist-packages (from requests->torch-geometric)
    (2024.2.2)
    Requirement already satisfied: joblib>=1.1.1 in /usr/local/lib/python3.10/dist-
    packages (from scikit-learn->torch-geometric) (1.3.2)
    Requirement already satisfied: threadpoolctl>=2.0.0 in
    /usr/local/lib/python3.10/dist-packages (from scikit-learn->torch-geometric)
    (3.3.0)
    Installing collected packages: torch-geometric
    Successfully installed torch-geometric-2.5.1
[2]: from google.colab import drive
     drive.mount('/content/drive')
    Mounted at /content/drive
    Loading chunks of the dataset
    Note: To preserve RAM, several unused variables are deleted throughout the norebook
[3]: import torch
     chunk1 = torch.load('/content/drive/MyDrive/SCI_data/first.pt')
[4]: chunk2 = torch.load('/content/drive/MyDrive/SCI_data/second (1).pt')
[9]: from torch.utils.data import Dataset
     class GraphDataset(Dataset):
       def __init__(self, data, preprocess):
         self.data = data
         self.preprocess = preprocess
       def __len__(self,):
         return len(self.data)
       def __getitem__(self, index):
         res = self.data[index]
         for p in self.preprocess:
           res = p(res)
         return res
```

```
chunk1+=chunk2
      del chunk2
[10]: from sklearn.model_selection import train_test_split
      rand_seed = 42
      X_train, X_test = train_test_split(chunk1, test_size=0.1, random_state =_
       →rand_seed)
      X_train, X_val = train_test_split(X_train, test_size=0.1, random_state =
       →rand_seed)
      print(len(X_train), len(X_val), len(X_val))
     14580 1620 1620
[12]: device = 'cuda'
[13]: from torch_geometric.loader import DataLoader
      train_loader = DataLoader(X_train, batch_size=32, shuffle=True)
      val_loader = DataLoader(X_val, batch_size=32, shuffle=False)
      test_loader = DataLoader(X_test, batch_size=32, shuffle=False)
[14]: !pip install torchmetrics
     Collecting torchmetrics
       Downloading torchmetrics-1.3.1-py3-none-any.whl (840 kB)
                                840.4/840.4
     kB 10.5 MB/s eta 0:00:00
     Requirement already satisfied: numpy>1.20.0 in
     /usr/local/lib/python3.10/dist-packages (from torchmetrics) (1.25.2)
     Requirement already satisfied: packaging>17.1 in /usr/local/lib/python3.10/dist-
     packages (from torchmetrics) (23.2)
     Requirement already satisfied: torch>=1.10.0 in /usr/local/lib/python3.10/dist-
     packages (from torchmetrics) (2.1.0+cu121)
     Collecting lightning-utilities>=0.8.0 (from torchmetrics)
       Downloading lightning_utilities-0.10.1-py3-none-any.whl (24 kB)
     Requirement already satisfied: setuptools in /usr/local/lib/python3.10/dist-
     packages (from lightning-utilities>=0.8.0->torchmetrics) (67.7.2)
     Requirement already satisfied: typing-extensions in
     /usr/local/lib/python3.10/dist-packages (from lightning-
     utilities>=0.8.0->torchmetrics) (4.10.0)
     Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-
     packages (from torch>=1.10.0->torchmetrics) (3.13.1)
     Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packages
     (from torch>=1.10.0->torchmetrics) (1.12)
     Requirement already satisfied: networkx in /usr/local/lib/python3.10/dist-
     packages (from torch>=1.10.0->torchmetrics) (3.2.1)
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Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages (from torch>=1.10.0->torchmetrics) (3.1.3)

Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from torch>=1.10.0->torchmetrics) (2023.6.0)

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Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->torch>=1.10.0->torchmetrics) (2.1.5)

Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dist-packages (from sympy->torch>=1.10.0->torchmetrics) (1.3.0)

Installing collected packages: lightning-utilities, torchmetrics

Successfully installed lightning-utilities-0.10.1 torchmetrics-1.3.1
```

```
[17]: import torch
      import torch.nn as nn
      import torch.nn.functional as F
      import torch.optim as optim
      from torch_geometric.nn import GCNConv, global_mean_pool, GATConv, SAGEConv, U
       SAGPooling
      from torch_geometric.nn.conv import GraphConv
      from torch_geometric.utils import to_undirected
      from torch_geometric.data import DataLoader
      from torchmetrics.classification import BinaryAUROC
      auroc = BinaryAUROC()
      class Network(nn.Module):
          def __init__(self, c_in, c_hidden, c_out, p=0.3):
              super().__init__()
              torch.manual_seed(123)
              self.conv1 = SAGEConv(c_in, c_hidden, aggr='mean')
              self.conv2 = SAGEConv(c_hidden, 3*c_hidden, aggr='mean')
              self.conv3 = SAGEConv(3*c_hidden, 2*c_hidden, aggr='mean')
              self.conv4 = SAGEConv(2*c_hidden, c_hidden, aggr='mean')
              # self.pool = SAGPooling(c_hidden)
              self.lin1 = nn.Linear(c_hidden, 4*c_out)
              self.lin2 = nn.Linear(4*c_out, c_out)
              self.p = p
          def forward(self, x, edge_index, batch, is_train):
              x = self.conv1(x, edge_index)
              x = x.relu()
              x = self.conv2(x, edge index)
```

```
x = x.relu()
        x = self.conv3(x, edge_index)
        x = x.relu()
        x = self.conv4(x, edge_index)
        x = global_mean_pool(x, batch)
        # classifier
        x = F.dropout(x, p=self.p, training=is_train)
        x = self.lin1(x)
        x = F.dropout(x, p=self.p, training=is_train)
        x = self.lin2(x)
        return x
def evaluate(loader):
   model.eval()
    total_loss = 0.0
    correct = 0
    total_samples = 0
    all_preds = []
    all_labels = []
    with torch.no_grad():
        for batch in loader:
            batch.to(device)
            # print(batch.edge_index)
            pred = model(batch.x.float(), batch.edge_index, batch.batch, False)
            target = F.one_hot(batch.y, 2).float()
            loss = criterion(pred, target)
            total_loss += loss.item()
            pred_labels = torch.softmax(pred, -1).argmax(dim=-1)
            correct += (pred_labels == batch.y).sum().item()
            total_samples += len(batch.y)
            all_labels.append(batch.y)
            all_preds.append(pred_labels)
    pred = all_preds[0]
    label = all_labels[0]
```

```
for p, l in zip(all_preds[1:], all_labels[1:]):
      pred = torch.cat([pred, p])
      label = torch.cat([label, 1])
    return total_loss / len(loader), correct / total_samples, auroc(pred, label)
# Training loop
num_epochs = 50
embedding_dim = 64
model = Network(c_in=5, c_hidden=embedding_dim, c_out=2).to(device)
optimizer = optim.Adam(model.parameters(), lr=3e-4)
criterion = nn.BCEWithLogitsLoss()
for epoch in range(num_epochs):
    model.train()
    epoch_loss = 0
    for idx, batch in enumerate(train_loader):
        batch = batch.to(device)
        pred = model(batch.x.float(), batch.edge_index, batch.batch, True)
        target = F.one_hot(batch.y, 2).float()
        loss = criterion(pred, target)
        epoch_loss += loss.item()
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()
    avg_train_loss = epoch_loss / len(train_loader)
    avg_val_loss, val_accuracy, val_auroc = evaluate(val_loader)
    print(f'Epoch {epoch + 1}/{num_epochs}, Train Loss: {avg_train_loss:.4f},_u
 →Val Loss: {avg_val_loss:.4f}, Val Accuracy: {val_accuracy:.4f}, Val AUROC:⊔

√{val_auroc:.4f}')

Epoch 1/50, Train Loss: 0.6614, Val Loss: 0.6176, Val Accuracy: 0.6796, Val
AUROC: 0.6809
Epoch 2/50, Train Loss: 0.6218, Val Loss: 0.6058, Val Accuracy: 0.6716, Val
AUROC: 0.6749
Epoch 3/50, Train Loss: 0.6071, Val Loss: 0.5873, Val Accuracy: 0.7136, Val
AUROC: 0.7143
Epoch 4/50, Train Loss: 0.6035, Val Loss: 0.5835, Val Accuracy: 0.7099, Val
AUROC: 0.7081
Epoch 5/50, Train Loss: 0.5970, Val Loss: 0.5874, Val Accuracy: 0.7154, Val
```

- AUROC: 0.7134
- Epoch 6/50, Train Loss: 0.5982, Val Loss: 0.5797, Val Accuracy: 0.7136, Val
- AUROC: 0.7123
- Epoch 7/50, Train Loss: 0.5988, Val Loss: 0.5819, Val Accuracy: 0.7179, Val
- AUROC: 0.7165
- Epoch 8/50, Train Loss: 0.5972, Val Loss: 0.5818, Val Accuracy: 0.7154, Val
- AUROC: 0.7150
- Epoch 9/50, Train Loss: 0.5964, Val Loss: 0.5774, Val Accuracy: 0.7198, Val
- AUROC: 0.7193
- Epoch 10/50, Train Loss: 0.5908, Val Loss: 0.5764, Val Accuracy: 0.7185, Val
- AUROC: 0.7179
- Epoch 11/50, Train Loss: 0.5922, Val Loss: 0.5772, Val Accuracy: 0.7142, Val
- AUROC: 0.7129
- Epoch 12/50, Train Loss: 0.5911, Val Loss: 0.5788, Val Accuracy: 0.7185, Val
- AUROC: 0.7165
- Epoch 13/50, Train Loss: 0.5915, Val Loss: 0.5765, Val Accuracy: 0.7198, Val
- AUROC: 0.7202
- Epoch 14/50, Train Loss: 0.5902, Val Loss: 0.5848, Val Accuracy: 0.7136, Val
- AUROC: 0.7152
- Epoch 15/50, Train Loss: 0.5913, Val Loss: 0.5774, Val Accuracy: 0.7235, Val
- AUROC: 0.7239
- Epoch 16/50, Train Loss: 0.5914, Val Loss: 0.5857, Val Accuracy: 0.7117, Val
- AUROC: 0.7134
- Epoch 17/50, Train Loss: 0.5910, Val Loss: 0.5784, Val Accuracy: 0.7235, Val
- AUROC: 0.7218
- Epoch 18/50, Train Loss: 0.5891, Val Loss: 0.5737, Val Accuracy: 0.7228, Val
- AUROC: 0.7229
- Epoch 19/50, Train Loss: 0.5907, Val Loss: 0.5722, Val Accuracy: 0.7185, Val
- AUROC: 0.7176
- Epoch 20/50, Train Loss: 0.5881, Val Loss: 0.5729, Val Accuracy: 0.7191, Val
- AUROC: 0.7179
- Epoch 21/50, Train Loss: 0.5878, Val Loss: 0.5750, Val Accuracy: 0.7228, Val
- AUROC: 0.7212
- Epoch 22/50, Train Loss: 0.5874, Val Loss: 0.5737, Val Accuracy: 0.7290, Val
- AUROC: 0.7285
- Epoch 23/50, Train Loss: 0.5871, Val Loss: 0.5815, Val Accuracy: 0.7142, Val
- AUROC: 0.7117
- Epoch 24/50, Train Loss: 0.5862, Val Loss: 0.5714, Val Accuracy: 0.7204, Val
- AUROC: 0.7201
- Epoch 25/50, Train Loss: 0.5862, Val Loss: 0.5728, Val Accuracy: 0.7222, Val
- AUROC: 0.7227
- Epoch 26/50, Train Loss: 0.5853, Val Loss: 0.5700, Val Accuracy: 0.7191, Val
- AUROC: 0.7189
- Epoch 27/50, Train Loss: 0.5873, Val Loss: 0.5689, Val Accuracy: 0.7222, Val
- AUROC: 0.7212
- Epoch 28/50, Train Loss: 0.5859, Val Loss: 0.5709, Val Accuracy: 0.7247, Val
- AUROC: 0.7235
- Epoch 29/50, Train Loss: 0.5907, Val Loss: 0.5702, Val Accuracy: 0.7210, Val

```
AUROC: 0.7196
Epoch 30/50, Train Loss: 0.5871, Val Loss: 0.5690, Val Accuracy: 0.7191, Val
AUROC: 0.7186
Epoch 31/50, Train Loss: 0.5860, Val Loss: 0.5742, Val Accuracy: 0.7241, Val
AUROC: 0.7227
Epoch 32/50, Train Loss: 0.5864, Val Loss: 0.5702, Val Accuracy: 0.7272, Val
AUROC: 0.7266
Epoch 33/50, Train Loss: 0.5848, Val Loss: 0.5683, Val Accuracy: 0.7290, Val
AUROC: 0.7279
Epoch 34/50, Train Loss: 0.5867, Val Loss: 0.5708, Val Accuracy: 0.7247, Val
AUROC: 0.7233
Epoch 35/50, Train Loss: 0.5839, Val Loss: 0.5757, Val Accuracy: 0.7185, Val
AUROC: 0.7194
Epoch 36/50, Train Loss: 0.5846, Val Loss: 0.5700, Val Accuracy: 0.7198, Val
AUROC: 0.7189
Epoch 37/50, Train Loss: 0.5854, Val Loss: 0.5772, Val Accuracy: 0.7265, Val
AUROC: 0.7257
Epoch 38/50, Train Loss: 0.5834, Val Loss: 0.5700, Val Accuracy: 0.7204, Val
AUROC: 0.7204
Epoch 39/50, Train Loss: 0.5855, Val Loss: 0.5707, Val Accuracy: 0.7265, Val
AUROC: 0.7264
Epoch 40/50, Train Loss: 0.5846, Val Loss: 0.5674, Val Accuracy: 0.7265, Val
AUROC: 0.7256
Epoch 41/50, Train Loss: 0.5834, Val Loss: 0.5676, Val Accuracy: 0.7259, Val
AUROC: 0.7249
Epoch 42/50, Train Loss: 0.5872, Val Loss: 0.5739, Val Accuracy: 0.7210, Val
AUROC: 0.7204
Epoch 43/50, Train Loss: 0.5837, Val Loss: 0.5696, Val Accuracy: 0.7247, Val
AUROC: 0.7248
Epoch 44/50, Train Loss: 0.5855, Val Loss: 0.5688, Val Accuracy: 0.7259, Val
AUROC: 0.7261
Epoch 45/50, Train Loss: 0.5859, Val Loss: 0.5711, Val Accuracy: 0.7204, Val
AUROC: 0.7184
Epoch 46/50, Train Loss: 0.5841, Val Loss: 0.5697, Val Accuracy: 0.7259, Val
AUROC: 0.7264
Epoch 47/50, Train Loss: 0.5845, Val Loss: 0.5666, Val Accuracy: 0.7278, Val
AUROC: 0.7268
Epoch 48/50, Train Loss: 0.5830, Val Loss: 0.5729, Val Accuracy: 0.7210, Val
AUROC: 0.7205
Epoch 49/50, Train Loss: 0.5826, Val Loss: 0.5708, Val Accuracy: 0.7259, Val
AUROC: 0.7264
Epoch 50/50, Train Loss: 0.5831, Val Loss: 0.5683, Val Accuracy: 0.7222, Val
AUROC: 0.7222
```

[20]: # Testing

test_loss, test_accuracy, test_auroc = evaluate(test_loader)

```
print(f'Test Loss: {test_loss:.4f}, Test Accuracy: {100*test_accuracy:.4f}%, □ 

→Test AUROC: {test_auroc:.4f}')
```

Test Loss: 0.5703, Test Accuracy: 72.3333%, Test AUROC: 0.7239

[]: