# Лабораторная работа №6:

"Разработка системы предсказания поведения на основании графовых моделей"

Цель: обучение работе с графовым типом данных и графовыми нейронными сетями.

Задача: подготовить графовый датасет из базы данных о покупках и построить модель предсказания совершения покупки.

## Графовые нейронные сети

**Графовые нейронные сети** - тип нейронной сети, которая напрямую работает со структурой графа. Типичным применениями GNN являются:

- Классификация узлов;
- Предсказание связей;
- Графовая классификация;
- Распознавание движений;
- Рекомендательные системы

В данной лабораторной работе будет происходить работа над **графовыми сверточными сетями**. Отличаются они от сверточных нейронных сетей нефиксированной структурой, функция свертки не является.

Подробнее можно прочитать тут: https://towardsdatascience.com/understanding-graph-convolutional-networks-for-node-classification-a2bfdb7aba7b

Тут можно почитать современные подходы к использованию графовых сверточных сетей https://paperswithcode.com/method/gcn

### Датасет

В качестве базы данных предлагаем использовать датасет о покупках пользователей в одном магазине товаров RecSys Challenge 2015 (https://www.kaggle.com/datasets/chadgostopp/recsys-challenge-2015).

Скачать датасет можно отсюда: https://drive.google.com/drive/folders/1gtAeXPTj-c0RwVOKreMrZ3bfSmCwl2y-?usp=sharing (lite-версия является облегченной версией исходного датасета, рекомендуем использовать её)

Также рекомендуем загружать данные в виде архива и распаковывать через пакет zipfile или/и скачивать датасет в собственный Google Drive и примонтировать его в колаб.

#### Установка библиотек, выгрузка исходных датасетов

```
# Slow method of installing pytorch geometric
# !pip install torch geometric
# !pip install torch sparse
# !pip install torch_scatter
# Install pytorch geometric
 !pip install torch-sparse -f https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
 !pip install torch-cluster -f https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
!pip install torch-spline-conv -f https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
 !pip install torch-geometric -f https://pytorch-geometric.com/whl/torch-1.11.0%2Bcul13.html
!pip install torch-scatter==2.0.8 -f https://data.pyg.org/whl/torch-1.11.0%2Bcu113.html
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Looking in links: https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
Collecting torch-sparse
  Downloading https://data.pyg.org/whl/torch-1.11.0%2Bcu113/torch_sparse-0.6.13-cp37-cp37m-linux_x86_64.whl (3.5 M
                                     | 3.5 MB 15.8 MB/s
Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from torch-sparse) (1.4.1)
Requirement already satisfied: numpy>=1.13.3 in /usr/local/lib/python3.7/dist-packages (from scipy->torch-sparse)
(1.21.6)
Installing collected packages: torch-sparse
Successfully installed torch-sparse-0.6.13
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Looking in links: https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
Collecting torch-cluster
 Downloading https://data.pyg.org/whl/torch-1.11.0%2Bcu113/torch_cluster-1.6.0-cp37-cp37m-linux_x86_64.whl (2.5 M
B)
```

```
Installing collected packages: torch-cluster
        Successfully installed torch-cluster-1.6.0
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        Looking in links: https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
        Collecting torch-spline-conv
          Downloading https://data.pyg.org/whl/torch-1.11.0%2Bcu113/torch spline conv-1.2.1-cp37-cp37m-linux x86 64.whl (7
        50 kB)
                                             | 750 kB 24.1 MB/s
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        Looking in links: https://pytorch-geometric.com/whl/torch-1.11.0%2Bcu113.html
        Collecting torch-geometric
          Downloading torch_geometric-2.0.4.tar.gz (407 kB)
                                              | 407 kB 34.7 MB/s
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        Requirement already satisfied: numpy in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (1.21.6)
        Requirement already satisfied: scipy in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (1.4.1)
        Requirement already satisfied: pandas in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (1.3.5)
        Requirement already satisfied: jinja2 in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (2.11.3) Requirement already satisfied: requests in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (2.23.0)
        Requirement already satisfied: pyparsing in /usr/local/lib/python3.7/dist-packages (from torch-geometric) (3.0.9)
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        2)
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        Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.7/dist-packages (from pandas->torch-geometri
        c) (2022.1)
        Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python3.7/dist-packages (from pandas->torc
        h-geometric) (2.8.2)
        Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.7/dist-packages (from python-dateutil>=2.7.3->pa
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        Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packages (from requests->torch-g
        eometric) (3.0.4)
        Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packages (from requests->torch-
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        Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1 in /usr/local/lib/python3.7/dist-packages (
        from requests->torch-geometric) (1.24.3)
        Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-packages (from requests->torch-geomet
        ric) (2.10)
        Requirement already satisfied: joblib>=0.11 in /usr/local/lib/python3.7/dist-packages (from scikit-learn->torch-ge
        ometric) (1.1.0)
        Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from scikit-learn->
        torch-geometric) (3.1.0)
        Building wheels for collected packages: torch-geometric
          Building wheel for torch-geometric (setup.py) ... done
          Created wheel for torch-geometric: filename=torch geometric-2.0.4-py3-none-any.whl size=616603 sha256=ccf1039bdb
        96b29de1113facbca43d5249d37dd2ebf013af4f2437580e6308df\\
          Stored in directory: /root/.cache/pip/wheels/18/a6/a4/ca18c3051fcead866fe7b85700ee2240d883562a1bc70ce421
        Successfully built torch-geometric
        Installing collected packages: torch-geometric
        Successfully installed torch-geometric-2.0.4
        Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
        Looking in links: https://data.pyg.org/whl/torch-1.11.0%2Bcu113.html
        Collecting torch-scatter==2.0.8
          Downloading torch_scatter-2.0.8.tar.gz (21 kB)
        Building wheels for collected packages: torch-scatter
          Building wheel for torch-scatter (setup.py) ... done
          Created wheel for torch-scatter: filename=torch scatter-2.0.8-cp37-cp37m-linux x86 64.whl size=3221894 sha256=e2
        541 a fcb 93936 fea 9 bc 9933d3 a 1 cecb 21db 57f8 508bf1b 432e8 3575a 15f2da 6\\
          Stored in directory: /root/.cache/pip/wheels/96/e4/4e/2bcc6de6a801960aedbca43f7106d268f766c3f9f8ab49b3a5
        Successfully built torch-scatter
        Installing collected packages: torch-scatter
        Successfully installed torch-scatter-2.0.8
In [2]: import numpy as np
         import pandas as pd
         import pickle
         import csv
         import os
         from sklearn.preprocessing import LabelEncoder
         import torch
         # PyG - PyTorch Geometric
         from torch_geometric.data import Data, DataLoader, InMemoryDataset
         from tadm import tadm
         RANDOM_SEED = 42 #@param { type: "integer" }
         BASE_DIR = '/content/' #@param { type: "string" }
```

np.random.seed(RANDOM\_SEED)

```
torch.cuda.is_available
 Out[3]: <function torch.cuda.is_available>
           # Unpack files from zip-file
           import zipfile
           with zipfile.ZipFile(BASE DIR + 'yoochoose-data-lite.zip', 'r') as zip ref:
               zip_ref.extractall(BASE DIR)
         Анализ исходных данных
 In [7]:
           # Read dataset of items in store
           df = pd.read_csv(BASE_DIR + 'yoochoose-clicks-lite.dat')
# df.columns = ['session_id', 'timestamp', 'item_id', 'category']
           df.head()
          /usr/local/lib/python3.7/dist-packages/IPython/core/interactiveshell.py:2882: DtypeWarning: Columns (3) have mixed
          types. Specify dtype option on import or set low_memory=False.
          exec(code_obj, self.user_global_ns, self.user_ns)
 Out[7]:
            session id
                                  timestamp
                                               item_id category
                    9 2014-04-06T11:26:24.127Z 214576500
          1
                    9 2014-04-06T11:28:54.654Z 214576500
                                                            0
          2
                    9 2014-04-06T11:29:13.479Z 214576500
                                                            0
          3
                   19 2014-04-01T20:52:12.357Z 214561790
                                                            0
                   19 2014-04-01T20:52:13.758Z 214561790
                                                             0
 In [8]:
           # Read dataset of purchases
           buy_df = pd.read_csv(BASE_DIR + 'yoochoose-buys-lite.dat')
           # buy_df.columns = ['session_id', 'timestamp', 'item_id', 'price', 'quantity']
           buy_df.head()
 Out[8]:
            session_id
                                   timestamp
                                               item_id price quantity
                420374 2014-04-06T18:44:58.314Z 214537888 12462
                420374 2014-04-06T18:44:58.325Z 214537850 10471
                489758 2014-04-06T09:59:52.422Z 214826955
          2
                                                        1360
                                                                   2
                489758 2014-04-06T09:59:52.476Z 214826715
                                                        732
                489758 2014-04-06T09:59:52.578Z 214827026
                                                       1046
                                                                   1
 In [9]: # Filter out item session with length < 2</pre>
           df['valid session'] = df.session id.map(df.groupby('session id')['item id'].size() > 2)
           df = df.loc[df.valid session].drop('valid session',axis=1)
           df.nunique()
 Out[9]: session_id
                         1000000
                         5557758
          timestamp
                           37644
          item id
                             275
          category
          dtype: int64
In [18]: # Randomly sample a couple of them
           NUM_SESSIONS = 50000 #@param { type: "integer" }
           sampled_session_id = np.random.choice(df.session_id.unique(), NUM_SESSIONS, replace=False)
           df = df.loc[df.session_id.isin(sampled_session_id)]
           df.nunique()
                          50000
Out[10]: session_id
                         278442
          timestamp
                          18461
          item id
          category
                            110
          dtype: int64
           # Average length of session
           df.groupby('session_id')['item_id'].size().mean()
Out[11]: 5.56902
In [12]: | # Encode item and category id in item dataset so that ids will be in range (0,len(df.item.unique()))
           item_encoder = LabelEncoder()
           category encoder = LabelEncoder()
           df['item_id'] = item_encoder.fit_transform(df.item_id)
           df['category']= category_encoder.fit_transform(df.category.apply(str))
           df.head()
              session id
                                    timestamp item_id category
            0
                      9 2014-04-06T11:26:24.127Z
                                                 3496
                                                            0
                      9 2014-04-06T11:28:54.654Z
            1
                                                 3496
                                                            0
```

```
103
                   171 2014-04-03T17:45:33.177Z 10137
                                                          0
In [13]:
          # Encode item and category id in purchase dataset
          buy_df = buy_df.loc[buy_df.session_id.isin(df.session_id)]
          buy_df['item_id'] = item_encoder.transform(buy_df.item_id)
          buy_df.head()
         /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#return
         ing-a-view-versus-a-copy
         This is separate from the ipykernel package so we can avoid doing imports until
                                  timestamp item_id
Out[13]:
             session id
                                                    price quantity
         46
                489491 2014-04-06T12:41:34.047Z
                                             12633
                                                    1046
         47
                489491 2014-04-06T12:41:34.091Z
                                             12634
                                                     627
                 70353 2014-04-06T10:55:06.086Z 14345 41783
         61
                489671 2014-04-03T15:48:37.392Z 12489
          62
                                                    4188
          63
                489671 2014-04-03T15:59:35.495Z 12489
                                                   4188
In [14]:
          # Get item dictionary with grouping by session
          buy_item_dict = dict(buy_df.groupby('session_id')['item_id'].apply(list))
          buy_item_dict
Out[14]: {714: [14720, 14915, 14917, 3089],
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2

102

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171 2014-04-03T17:45:25.575Z

3496

10049

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```

#### Сборка выборки для обучения

```
In [5]: # Transform df into tensor data
                      def transform_dataset(df, buy_item_dict):
                                data_list = []
                                # Group by session
                                grouped = df.groupby('session_id')
                                for session_id, group in tqdm(grouped):
                                          le = LabelEncoder()
                                          sess_item_id = le.fit_transform(group.item_id)
                                         group = group.reset_index(drop=True)
group['sess_item_id'] = sess_item_id
                                          #get input features
                                         node_features = group.loc[group.session_id==session_id,
                                                                                                               ['sess_item_id','item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')[['item_id','category']]].sort_values('sess_item_id','category')
                                          node_features = torch.LongTensor(node_features).unsqueeze(1)
                                          target_nodes = group.sess_item_id.values[1:]
                                          source_nodes = group.sess_item_id.values[:-1]
                                         edge_index = torch.tensor([source_nodes,
                                                                                                    target_nodes], dtype=torch.long)
                                         x = node features
                                          #get result
                                          if session_id in buy_item_dict:
                                                    positive_indices = le.transform(buy_item_dict[session_id])
                                                    label = np.zeros(len(node_features))
                                                    label[positive indices] = 1
                                          else:
                                                   label = [0] * len(node_features)
                                         y = torch.FloatTensor(label)
                                         data = Data(x=x, edge index=edge index, y=y)
                                          data_list.append(data)
                                return data list
                      # Pytorch class for creating datasets
                      class YooChooseDataset(InMemoryDataset):
                                         __init__(self, root, transform=None, pre_transform=None):
super(YooChooseDataset, self).__init__(root, transform, pre_transform)
self.data, self.slices = torch.load(self.processed_paths[0])
                                @property
                                def raw_file_names(self):
                                         return []
                                @property
                                def processed_file_names(self):
                                          return [BASE_DIR+'yoochoose click binary 100000 sess.dataset']
                                def download(self):
                                          pass
                                def process(self):
                                          data_list = transform_dataset(df, buy_item_dict)
                                          data, slices = self.collate(data_list)
                                          torch.save((data, slices), self.processed_paths[0])
```

#### Разделение выборки

```
In [ ]:
In [16]: # train_test_split
                               dataset = dataset.shuffle()
                              one tenth length = int(len(dataset) * 0.1)
                               train_dataset = dataset[:one_tenth_length * 8]
                               val_dataset = dataset[one_tenth_length*8:one_tenth_length * 9]
                               test_dataset = dataset[one_tenth_length*9:]
                              len(train dataset), len(val dataset), len(test dataset)
Out[16]: (40000, 5000, 5000)
In [17]: train_dataset
Out[17]: YooChooseDataset(40000)
In [18]: # Load dataset into PyG loaders
                              batch size= 512
                              train_loader = DataLoader(train_dataset, batch_size=batch_size)
                               val loader = DataLoader(val dataset, batch size=batch size)
                              test loader = DataLoader(test dataset, batch size=batch size)
                             /usr/local/lib/python 3.7/dist-packages/torch\_geometric/deprecation.py: 12: UserWarning: 'data.DataLoader' is deprecation.py: 13: UserWarning: 'data.DataL
                            ated, use 'loader.DataLoader' instead
                               warnings.warn(out)
 In [19]: # Load dataset into PyG loaders
                              num items = df.item id.max() +1
                              num_categories = df.category.max()+1
                              num_items , num_categories
Out[19]: (18461, 109)
```

#### Настройка модели для обучения

```
In [53]:
         embed dim = 128
          from Torch_geometric.nn import GraphConv, TopKPooling, GatedGraphConv, SAGEConv, SGConv
          from torch geometric.nn import global mean pool as gap, global max pool as gmp
          import torch.nn.functional as F
          class Net(torch.nn.Module):
                   init__(self);
              def
                  super(Net, self).
                                    __init__()
                  # Model Structure
                  self.conv1 = GraphConv(embed_dim * 2, 128)
                  self.pool1 = TopKPooling(128, ratio=0.9)
                  self.conv2 = GraphConv(128, 128)
                  self.pool2 = TopKPooling(128, ratio=0.9)
                  self.conv3 = GraphConv(128, 128)
                  self.pool3 = TopKPooling(128, ratio=0.9)
                  self.item_embedding = torch.nn.Embedding(num_embeddings=num_items, embedding_dim=embed_dim)
                  \tt self.category\_embedding = torch.nn.Embedding(num\_embeddings=num\_categories, \ embedding\_dim=embed\_dim)
                  self.lin1 = torch.nn.Linear(256, 256)
                  self.lin2 = torch.nn.Linear(256,
                  self.bn1 = torch.nn.BatchNorm1d(128)
                  self.bn2 = torch.nn.BatchNorm1d(64)
                  self.act1 = torch.nn.ReLU()
                  self.act2 = torch.nn.ReLU()
              # Forward step of a model
              def forward(self, data):
                  x, edge index, batch = data.x, data.edge index, data.batch
                  item_id = x[:,:,0]
                  category = x[:,:,1]
                  emb_item = self.item_embedding(item_id).squeeze(1)
                  emb_category = self.category_embedding(category).squeeze(1)
                  x = torch.cat([emb_item, emb_category], dim=1)
                  x = F.relu(self.conv1(x, edge index))
                  # print(x.shape)
                  r = self.pool1(x, edge_index, None, batch)
```

```
# print(r)
x, edge_index, _, batch, _, _ = self.pool1(x, edge_index, None, batch)
x1 = torch.cat([gmp(x, batch), gap(x, batch)], dim=1)
x = F.relu(self.conv2(x, edge index))
x, edge_index, _, batch, _, _ = self.pool2(x, edge_index, None, batch) x2 = torch.cat([gmp(x, batch), gap(x, batch)], dim=1)
x = F.relu(self.conv3(x, edge index))
x, edge_index, _, batch, _, _ = self.pool3(x, edge_index, None, batch) x3 = torch.cat([gmp(x, batch), gap(x, batch)], dim=1)
x = x1 + x2 + x3
x = self.lin1(x)
x = self.act1(x)
x = self.lin2(x)
x = F.dropout(x, p=0.13, training=self.training)
x = self.act2(x)
outputs = []
for i in range(x.size(0)):
     output = torch.matmul(emb_item[data.batch == i], x[i,:])
     outputs.append(output)
x = torch.cat(outputs, dim=0)
x = torch.sigmoid(x)
return x
```

#### Обучение нейронной сверточной сети

```
In [54]: # Enable CUDA computing
          device = torch.device('cuda')
          model = Net().to(device)
          # Choose optimizer and criterion for learning
          optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
          crit = torch.nn.BCELoss()
In [55]: # Train function
          def train():
               model.train()
               loss all = 0
               for data in train loader:
                   data = data.to(device)
                   optimizer.zero_grad()
                   output = model(data)
                   label = data.y.to(device)
loss = crit(output, label)
                   loss.backward()
                   loss all += data.num graphs * loss.item()
                   optimizer.step()
               return loss_all / len(train_dataset)
In [56]:
          # Evaluate result of a model
          from sklearn.metrics import roc_auc_score
          def evaluate(loader):
```

```
model.eval()
predictions = []
labels = []
with torch.no_grad():
    for data in loader:
        data = data.to(device)
        pred = model(data).detach().cpu().numpy()
        label = data.y.detach().cpu().numpy()
        predictions.append(pred)
        labels.append(label)
predictions = np.hstack(predictions)
labels = np.hstack(labels)
return roc_auc_score(labels, predictions)
```

```
# Train a model
In [57]:
          NUM_EPOCHS = 15 #@param { type: "integer" }
          for epoch in tqdm(range(NUM_EPOCHS)):
              loss = train()
              train acc = evaluate(train loader)
              val acc = evaluate(val loader)
              test_acc = evaluate(test_loader)
```

```
7%|
                       | 1/15 [00:48<11:21, 48.69s/it]
         Epoch: 000, Loss: 0.63157, Train Auc: 0.51078, Val Auc: 0.51656, Test Auc: 0.52349
                       | 2/15 [01:27<09:19, 43.02s/it]
         Epoch: 001, Loss: 0.38785, Train Auc: 0.58842, Val Auc: 0.57330, Test Auc: 0.56757
          20%|
                        | 3/15 [02:06<08:15, 41.27s/it]
         Epoch: 002, Loss: 0.30027, Train Auc: 0.63347, Val Auc: 0.59707, Test Auc: 0.59547
         27%|
                       | 4/15 [02:45<07:23, 40.29s/it]
         Epoch: 003, Loss: 0.27669, Train Auc: 0.68141, Val Auc: 0.60854, Test Auc: 0.61198
          33%|
                       | 5/15 [03:24<06:36, 39.64s/it]
         Epoch: 004, Loss: 0.24435, Train Auc: 0.71573, Val Auc: 0.62193, Test Auc: 0.61930
          40%|
                       | 6/15 [04:03<05:54, 39.37s/it]
         Epoch: 005, Loss: 0.23586, Train Auc: 0.76196, Val Auc: 0.62017, Test Auc: 0.63280
                      | 7/15 [04:41<05:12, 39.08s/it]
         Epoch: 006, Loss: 0.20968, Train Auc: 0.79929, Val Auc: 0.63024, Test Auc: 0.63932
          53%|
                      | 8/15 [05:19<04:31, 38.79s/it]
         Epoch: 007, Loss: 0.19800, Train Auc: 0.82909, Val Auc: 0.63727, Test Auc: 0.64306
          60%| | 9/15 [05:58<03:52, 38.69s/it]
         Epoch: 008, Loss: 0.18686, Train Auc: 0.85950, Val Auc: 0.63802, Test Auc: 0.64612
          67%| | 10/15 [06:36<03:13, 38.64s/it]
         Epoch: 009, Loss: 0.17055, Train Auc: 0.88926, Val Auc: 0.64641, Test Auc: 0.65143
                   | 11/15 [07:14<02:33, 38.45s/it]
         Epoch: 010, Loss: 0.15698, Train Auc: 0.89951, Val Auc: 0.65138, Test Auc: 0.66006
                   | 12/15 [07:52<01:55, 38.39s/it]
         Epoch: 011, Loss: 0.14569, Train Auc: 0.92544, Val Auc: 0.64994, Test Auc: 0.65996
                    | 13/15 [08:31<01:16, 38.28s/it]
         Epoch: 012, Loss: 0.13154, Train Auc: 0.94664, Val Auc: 0.66301, Test Auc: 0.66276
                     | 14/15 [09:08<00:38, 38.19s/it]
         Epoch: 013, Loss: 0.12013, Train Auc: 0.95590, Val Auc: 0.65600, Test Auc: 0.66621
                     | 15/15 [09:47<00:00, 39.15s/it]
         Epoch: 014, Loss: 0.11669, Train Auc: 0.96066, Val Auc: 0.65418, Test Auc: 0.65919
        Проверка результата с помощью примеров
In [25]: # Подход №1 - из датасета
          evaluate(DataLoader(test dataset[40:60], batch size=10))
         /usr/local/lib/python3.7/dist-packages/torch_geometric/deprecation.py:12: UserWarning: 'data.DataLoader' is deprec
         ated, use 'loader.DataLoader' instead
         warnings.warn(out)
Out[25]: 0.7456790123456791
         # Подход №2 - через создание сессии покупок
In [26]:
          test_df = pd.DataFrame([
                [-1, 15219, 0],
                [-1, 15431, 0],
                [-1, 14371, 0],
                [-1, 15745, 0],
                [-2, 14594, 0],
                [-2, 16972, 11],
                [-2, 16943, 0],
                [-3, 17284, 0]
          ], columns=['session id', 'item id', 'category'])
          test_data = transform_dataset(test_df, buy_item_dict)
          test_data = DataLoader(test_data, batch_size=1)
          with torch.no grad():
              model eval()
              for data in test data:
                  data = data.to(device)
                 pred = model(data).detach().cpu().numpy()
                  print(data, pred)
                   | 3/3 [00:00<00:00, 175.68it/s]
         DataBatch(x=[1, 1, 2], edge_index=[2, 0], y=[1], batch=[1], ptr=[2]) [0.0861028]
         DataBatch(x=[3, 1, 2], edge_index=[2, 2], y=[3], batch=[3], ptr=[2]) [0.01035642 0.09229142 0.01806829]
         DataBatch(x=[4, 1, 2], edge_index=[2, 3], y=[4], batch=[4], ptr=[2]) [0.23247197 0.6972481 0.06574864 0.06372362]
         /usr/local/lib/python3.7/dist-packages/torch_geometric/deprecation.py:12: UserWarning: 'data.DataLoader' is deprec
         ated, use 'loader.DataLoader' instead
         warnings.warn(out)
```

print('Epoch: {:03d}, Loss: {:.5f}, Train Auc: {:.5f}, Val Auc: {:.5f}, Test Auc: {:.5f}'.

format(epoch, loss, train\_acc, val\_acc, test\_acc))