

Фреймворк PyTorch для разработки искусственных нейронных сетей

Урок 8. Generative Adversarial Networks

Практическое задание

Домашнее задание к уроку 8

Переписать загрузку данных с python функций на Dataset и Dataloader и применить сеть с attention

Выполнил **Соковнин ИЛ**

```
B [1]: import torch
import torch.nn as nn
import torch.nn.functional as F

import numpy as np
import pandas as pd
```

```
B [2]: epochs = 30
num_samples = 10000
data_path = './data/fra-eng/fra.txt'
```

```
B [3]: class eng_fra_dataset(torch.utils.data.Dataset):
    def __init__(self, file_name=data_path, num_samples=num_samples, nrow=num_samples):

        input_vocab = set()
        output_vocab = set()

        print(f'Загружаем файл: {file_name}')
        df = pd.read_csv(data_path, sep = '\t', header = None, nrow=num_samples)
        df.rename(columns = {0 : 'eng', 1: 'fra'}, inplace = True)
        # df = df[['eng', 'fra']]

        input_texts = list(df['eng'])
        target_texts = list(df['fra'])
        texts = df[['eng', 'fra']].values.tolist()

        for input_text in input_texts:
            for word in input_text.split():
                input_vocab.add(word.strip())

        for target_text in target_texts:
            for word in target_text.split():
                output_vocab.add(word.strip())

        output_vocab = set(list(output_vocab))

        input_vocab2index = {word: i+2 for i, word in enumerate(input_vocab)}
        output_vocab2index = {word: i+2 for i, word in enumerate(output_vocab)}

        self.texts = texts
        self.input_texts = input_texts
        self.target_texts = target_texts
        self.input_vocab2index = input_vocab2index
        self.output_vocab2index = output_vocab2index

        print('Файл {file_name} загружен')

    def __len__(self):
        return len(self.input_vocab)

    def __getitem__(self, idx):
        return self.texts[idx]

    def print_ds(self):
        for i in self.input_vocab2index:
            if self.input_vocab2index[i] < 12:
                print(i, self.input_vocab2index[i])
```

```

B [4]: def indexesFromSentence(sentence, vocab):
        return [vocab.get(word.strip(), 0) for word in sentence.split(' ')]

def tensorFromSentence(sentence, vocab):
    indexes = indexesFromSentence(sentence, vocab)
    indexes.append(1)
    return torch.tensor(indexes, dtype=torch.long).view(-1, 1)

def tensorsFromSent(input_sentences, output_sentences, input_vocab2index, output_vocab2index):
    input_tensor = tensorFromSentence(input_sentences, input_vocab2index)
    target_tensor = tensorFromSentence(output_sentences, output_vocab2index)
    return (input_tensor, target_tensor)

```

```

B [5]: class EncoderRNN(nn.Module):
        def __init__(self, input_size, hidden_size):
            super(EncoderRNN, self).__init__()
            self.hidden_size = hidden_size

            self.embedding = nn.Embedding(input_size, hidden_size)
            self.gru = nn.GRU(hidden_size, hidden_size)

        def forward(self, input, hidden):
            embedded = self.embedding(input).view(1, 1, -1)
            output = embedded
            output, hidden = self.gru(output, hidden)
            return output, hidden

        def initHidden(self):
            return torch.zeros(1, 1, self.hidden_size)

class AttnDecoderRNN(nn.Module):
    def __init__(self, hidden_size, output_size, dropout_p=0.1, max_length=10):
        super(AttnDecoderRNN, self).__init__()
        self.hidden_size = hidden_size
        self.output_size = output_size
        self.dropout_p = dropout_p
        self.max_length = max_length

        self.embedding = nn.Embedding(self.output_size, self.hidden_size)
        self.attn = nn.Linear(self.hidden_size * 2, self.max_length)
        self.attn_combine = nn.Linear(self.hidden_size * 2, self.hidden_size)
        self.dropout = nn.Dropout(self.dropout_p)
        self.gru = nn.GRU(self.hidden_size, self.hidden_size)
        self.out = nn.Linear(self.hidden_size, self.output_size)

    def forward(self, input, hidden, encoder_outputs):
        embedded = self.embedding(input).view(1, 1, -1)
        embedded = self.dropout(embedded)

        attn_weights = F.softmax(
            self.attn(torch.cat((embedded[0], hidden[0]), 1)), dim=-1)
        attn_applied = torch.bmm(attn_weights.unsqueeze(0),
                                encoder_outputs.unsqueeze(0))

        output = torch.cat((embedded[0], attn_applied[0]), 1)
        output = self.attn_combine(output).unsqueeze(0)

        #output = F.relu(output)
        output, hidden = self.gru(output, hidden)

        output = F.log_softmax(self.out(output[0]), dim=-1)
        return output, hidden, attn_weights

    def initHidden(self):
        return torch.zeros(1, 1, self.hidden_size)

```

```

B [6]: def train(input_tensor, target_tensor, encoder, decoder, encoder_optimizer, decoder_optimizer, criterion, max_length=10):
    encoder_hidden = encoder.initHidden()

    encoder_optimizer.zero_grad()
    decoder_optimizer.zero_grad()

    input_length = input_tensor.size(0)
    target_length = target_tensor.size(0)

    encoder_outputs = torch.zeros(max_length, encoder.hidden_size)

    loss = 0

    for ei in range(input_length):
        encoder_output, encoder_hidden = encoder(
            input_tensor[ei], encoder_hidden)
        encoder_outputs[ei] = encoder_output[0, 0]

    decoder_input = torch.tensor([[0]])

    decoder_hidden = encoder_hidden

    for di in range(target_length):
        decoder_output, decoder_hidden, decoder_attention = decoder(
            decoder_input, decoder_hidden, encoder_outputs)
        topv, topi = decoder_output.topk(1)
        decoder_input = topi.squeeze().detach() # detach from history as input

        loss += criterion(decoder_output, target_tensor[di])
        if decoder_input.item() == 1:
            break

    loss.backward()

    encoder_optimizer.step()
    decoder_optimizer.step()

    return loss.item() / target_length

```

```

B [7]: ds = eng_fra_dataset()
      # ds.print_ds()
      ds[1]

```

Загружаем файл: ./data/fra-eng/fra.txt
 Файл {file_name} загружен

```
Out[7]: ['Hi.', 'Salut !']
```

```

B [8]: encoder = EncoderRNN(len(ds.input_vocab2index)+2, 30)
        attn_decoder1 = AttnDecoderRNN(30, len(ds.output_vocab2index)+2, dropout_p=0.1)

        encoder_optimizer = torch.optim.SGD(encoder.parameters(), lr=0.01)
        decoder_optimizer = torch.optim.SGD(attn_decoder1.parameters(), lr=0.01)
        training_pairs = np.random.randint(0, len(ds.input_texts), size=10000)
        criterion = nn.NLLLoss()

        print_loss_total = 0

        for i in range(1000):
            input_tensor, target_tensor = tensorsFromSent(ds.input_texts[training_pairs[i]], ds.target_texts[training_pairs[i]],
                                                            ds.input_vocab2index, ds.output_vocab2index)

            loss = train(input_tensor, target_tensor, encoder,
                        attn_decoder1, encoder_optimizer, decoder_optimizer, criterion)
            print_loss_total += loss

            print_loss_avg = print_loss_total / 1
            print_loss_total = 0
            print('%d %d%%) %.4f' % (i, i / 10 * 100, print_loss_avg))

```

```

(0 0%) 8.6540
(1 10%) 8.6432
(2 20%) 8.7114
(3 30%) 8.7780
(4 40%) 8.6563
(5 50%) 8.6907
(6 60%) 8.6709
(7 70%) 8.6171
(8 80%) 8.7057
(9 90%) 8.6936
(10 100%) 8.5683
(11 110%) 8.5469
(12 120%) 8.7139
(13 130%) 8.5683
(14 140%) 8.5551
(15 150%) 8.5791
(16 160%) 8.7090
(17 170%) 8.6746
(18 180%) 8.5785
(19 190%) 8.5545
(20 200%) 8.5442
(21 210%) 8.6220
(22 220%) 8.4640
(23 229%) 8.4238
(24 240%) 8.4129
(25 250%) 8.6544
(26 260%) 8.5125
(27 270%) 8.6281
(28 280%) 8.4944
(29 290%) 8.5364
(30 300%) 4.3646
(31 310%) 2.8673
(32 320%) 3.4719
(33 330%) 5.6825
(34 340%) 6.9881
(35 350%) 3.6989
(36 360%) 5.7402
(37 370%) 8.4526
(38 380%) 6.3972
(39 390%) 2.8499
(40 400%) 5.6791
(41 409%) 3.5371
(42 420%) 6.9641
(43 430%) 3.4441
(44 440%) 2.9019
(45 450%) 3.5218
(46 459%) 8.5855
(47 470%) 3.4459
(48 480%) 3.3896
(49 490%) 5.7799
(50 500%) 8.3430
(51 509%) 4.1402
(52 520%) 8.3600
(53 530%) 3.4874
(54 540%) 3.4993
(55 550%) 5.6532
(56 560%) 8.3228
(57 570%) 8.2857
(58 580%) 3.3908
(59 590%) 3.4761
(60 600%) 2.9122
(61 610%) 5.7576
(62 620%) 2.1255
(63 630%) 2.8752
(64 640%) 3.4388

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(65 650%) 5.7602
(66 660%) 2.8140
(67 670%) 4.3067
(68 680%) 2.7792
(69 690%) 4.2024
(70 700%) 3.4899
(71 710%) 4.3297
(72 720%) 4.2798
(73 730%) 4.1941
(74 740%) 1.7677
(75 750%) 3.3655
(76 760%) 5.6881
(77 770%) 3.4872
(78 780%) 2.8212
(79 790%) 3.4471
(80 800%) 2.8373
(81 810%) 2.8392
(82 819%) 1.4405
(83 830%) 4.3065
(84 840%) 3.4888
(85 850%) 2.1211
(86 860%) 1.7083
(87 869%) 5.7449
(88 880%) 4.3728
(89 890%) 4.3166
(90 900%) 5.6929
(91 910%) 5.7073
(92 919%) 3.3514
(93 930%) 4.2191
(94 940%) 5.8076
(95 950%) 4.3092
(96 960%) 3.3367
(97 969%) 3.4265
(98 980%) 4.1652
(99 990%) 4.3131
(100 1000%) 3.3042
(101 1010%) 4.2407
(102 1019%) 4.3077
(103 1030%) 4.0787
(104 1040%) 3.4712
(105 1050%) 6.5220
(106 1060%) 5.4730
(107 1070%) 4.3562
(108 1080%) 3.3208
(109 1090%) 3.3962
(110 1100%) 2.6828
(111 1110%) 3.4143
(112 1120%) 5.2917
(113 1130%) 3.3450
(114 1140%) 4.3646
(115 1150%) 8.1083
(116 1160%) 3.2043
(117 1170%) 8.2521
(118 1180%) 3.2850
(119 1190%) 6.1555
(120 1200%) 4.1405
(121 1210%) 3.1853
(122 1220%) 8.2854
(123 1230%) 3.4270
(124 1240%) 5.8434
(125 1250%) 3.6965
(126 1260%) 5.1788
(127 1270%) 6.3482
(128 1280%) 5.2209
(129 1290%) 4.3987
(130 1300%) 2.7631
(131 1310%) 4.9685
(132 1320%) 6.4926
(133 1330%) 6.6403
(134 1340%) 2.7461
(135 1350%) 6.4476
(136 1360%) 6.1024
(137 1370%) 2.7978
(138 1380%) 6.4351
(139 1390%) 4.8719
(140 1400%) 6.3706
(141 1410%) 5.2169
(142 1420%) 5.2958
(143 1430%) 8.3757
(144 1440%) 3.7746
(145 1450%) 5.1228
(146 1460%) 6.4771
(147 1470%) 6.2795
(148 1480%) 8.0541
(149 1490%) 4.1945
(150 1500%) 3.2948
(151 1510%) 3.3078
(152 1520%) 3.4067

(153 1530%) 3.3969
(154 1540%) 4.3285
(155 1550%) 4.1467
(156 1560%) 2.3881
(157 1570%) 6.6833
(158 1580%) 4.2498
(159 1590%) 5.7011
(160 1600%) 6.3458
(161 1610%) 8.0959
(162 1620%) 6.1270
(163 1630%) 4.4043
(164 1639%) 5.9884
(165 1650%) 6.2670
(166 1660%) 5.1088
(167 1670%) 7.8307
(168 1680%) 4.3583
(169 1689%) 5.0198
(170 1700%) 2.5263
(171 1710%) 2.4644
(172 1720%) 3.7579
(173 1730%) 3.8781
(174 1739%) 5.1875
(175 1750%) 5.2423
(176 1760%) 6.4581
(177 1770%) 6.3770
(178 1780%) 4.4186
(179 1789%) 8.1706
(180 1800%) 5.1959
(181 1810%) 6.2181
(182 1820%) 5.3593
(183 1830%) 6.2772
(184 1839%) 3.4379
(185 1850%) 5.4718
(186 1860%) 5.8691
(187 1870%) 4.1194
(188 1880%) 5.9407
(189 1889%) 6.2854
(190 1900%) 4.9510
(191 1910%) 4.9567
(192 1920%) 7.9553
(193 1930%) 6.4501
(194 1939%) 5.0072
(195 1950%) 7.9465
(196 1960%) 4.2768
(197 1970%) 7.6960
(198 1980%) 5.4163
(199 1989%) 2.3627
(200 2000%) 3.3014
(201 2010%) 2.9093
(202 2020%) 4.2868
(203 2030%) 5.1238
(204 2039%) 3.4426
(205 2050%) 3.3029
(206 2060%) 3.6530
(207 2070%) 6.3795
(208 2080%) 8.2970
(209 2090%) 4.1029
(210 2100%) 3.0094
(211 2110%) 2.8075
(212 2120%) 5.7466
(213 2130%) 6.8209
(214 2140%) 4.0509
(215 2150%) 6.3899
(216 2160%) 6.6765
(217 2170%) 4.8609
(218 2180%) 5.7061
(219 2190%) 2.6665
(220 2200%) 4.2997
(221 2210%) 4.6779
(222 2220%) 7.7399
(223 2230%) 3.4904
(224 2240%) 5.9524
(225 2250%) 3.3732
(226 2260%) 6.5530
(227 2270%) 6.3897
(228 2280%) 4.3997
(229 2290%) 6.3131
(230 2300%) 6.1737
(231 2310%) 5.6638
(232 2320%) 6.5595
(233 2330%) 4.7846
(234 2340%) 6.2470
(235 2350%) 5.2470
(236 2360%) 5.0243
(237 2370%) 4.0766
(238 2380%) 4.9771
(239 2390%) 4.1620
(240 2400%) 4.2752

(241 2410%) 6.1546
(242 2420%) 8.0127
(243 2430%) 7.3944
(244 2440%) 3.7852
(245 2450%) 5.5639
(246 2460%) 5.8070
(247 2470%) 4.5081
(248 2480%) 3.3177
(249 2490%) 6.1424
(250 2500%) 4.8769
(251 2510%) 5.6767
(252 2520%) 6.3313
(253 2530%) 5.8503
(254 2540%) 5.1131
(255 2550%) 6.2380
(256 2560%) 6.6334
(257 2570%) 4.9057
(258 2580%) 7.5600
(259 2590%) 6.3421
(260 2600%) 6.0600
(261 2610%) 5.9970
(262 2620%) 7.5013
(263 2630%) 2.9166
(264 2640%) 4.4993
(265 2650%) 6.3387
(266 2660%) 7.5894
(267 2670%) 4.9674
(268 2680%) 4.9916
(269 2690%) 7.0191
(270 2700%) 3.1235
(271 2710%) 6.2449
(272 2720%) 4.3063
(273 2730%) 3.6311
(274 2740%) 5.1241
(275 2750%) 4.2560
(276 2760%) 6.5261
(277 2770%) 6.2576
(278 2780%) 7.5537
(279 2790%) 6.7295
(280 2800%) 3.4686
(281 2810%) 4.4341
(282 2820%) 4.6484
(283 2830%) 4.1926
(284 2840%) 5.6641
(285 2850%) 7.4276
(286 2860%) 6.3403
(287 2870%) 4.8021
(288 2880%) 6.3185
(289 2890%) 6.1308
(290 2900%) 4.9397
(291 2910%) 6.1703
(292 2920%) 5.1116
(293 2930%) 5.1514
(294 2940%) 7.4930
(295 2950%) 6.1430
(296 2960%) 5.2488
(297 2970%) 6.5583
(298 2980%) 6.4585
(299 2990%) 2.5360
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(301 3010%) 5.3094
(302 3020%) 4.7731
(303 3030%) 3.8094
(304 3040%) 6.4413
(305 3050%) 4.8452
(306 3060%) 4.8443
(307 3070%) 6.7259
(308 3080%) 5.0113
(309 3090%) 4.2529
(310 3100%) 5.3686
(311 3110%) 3.5010
(312 3120%) 7.3298
(313 3130%) 3.0126
(314 3140%) 5.7961
(315 3150%) 4.0713
(316 3160%) 4.1895
(317 3170%) 4.2154
(318 3180%) 3.3621
(319 3190%) 3.7719
(320 3200%) 3.5461
(321 3210%) 3.8672
(322 3220%) 2.2592
(323 3229%) 4.4236
(324 3240%) 4.1651
(325 3250%) 4.9972
(326 3260%) 6.3210
(327 3270%) 5.6646
(328 3279%) 3.1966

(329 3290%) 1.8386
(330 3300%) 6.2228
(331 3310%) 2.9484
(332 3320%) 2.9042
(333 3329%) 6.2112
(334 3340%) 5.7372
(335 3350%) 2.9380
(336 3360%) 2.9884
(337 3370%) 6.4366
(338 3379%) 4.8631
(339 3390%) 2.9941
(340 3400%) 5.9566
(341 3410%) 6.0075
(342 3420%) 2.9243
(343 3429%) 4.9600
(344 3440%) 6.1358
(345 3450%) 4.8024
(346 3460%) 6.3665
(347 3470%) 4.5617
(348 3479%) 5.5599
(349 3490%) 5.3879
(350 3500%) 2.4614
(351 3510%) 5.3554
(352 3520%) 4.6011
(353 3529%) 6.9401
(354 3540%) 6.3419
(355 3550%) 5.3728
(356 3560%) 7.0425
(357 3570%) 2.2512
(358 3579%) 2.2133
(359 3590%) 4.6224
(360 3600%) 4.7654
(361 3610%) 3.4882
(362 3620%) 6.1286
(363 3629%) 3.6462
(364 3640%) 3.8383
(365 3650%) 3.8151
(366 3660%) 5.7941
(367 3670%) 2.7073
(368 3679%) 4.1310
(369 3690%) 5.8564
(370 3700%) 4.9188
(371 3710%) 3.4872
(372 3720%) 2.7547
(373 3729%) 4.3752
(374 3740%) 4.2608
(375 3750%) 2.6504
(376 3760%) 4.4365
(377 3770%) 4.5438
(378 3779%) 4.0117
(379 3790%) 3.2426
(380 3800%) 2.8025
(381 3810%) 2.5820
(382 3820%) 4.7871
(383 3829%) 2.3473
(384 3840%) 3.3660
(385 3850%) 4.7917
(386 3860%) 6.2630
(387 3870%) 2.2825
(388 3879%) 5.9082
(389 3890%) 6.0612
(390 3900%) 6.6363
(391 3910%) 6.3408
(392 3920%) 3.3637
(393 3929%) 2.9790
(394 3940%) 1.2784
(395 3950%) 4.6737
(396 3960%) 4.6312
(397 3970%) 4.9905
(398 3979%) 4.4991
(399 3990%) 2.3875
(400 4000%) 5.5885
(401 4010%) 5.1912
(402 4020%) 5.1969
(403 4029%) 3.3273
(404 4040%) 3.9910
(405 4050%) 2.1268
(406 4060%) 6.0192
(407 4070%) 5.8745
(408 4079%) 6.8655
(409 4090%) 2.8233
(410 4100%) 4.5174
(411 4110%) 4.6348
(412 4120%) 5.2043
(413 4130%) 5.9632
(414 4140%) 3.6224
(415 4150%) 3.3544
(416 4160%) 4.4811

(417 4170%) 7.1022
(418 4180%) 2.6919
(419 4190%) 3.3021
(420 4200%) 6.3905
(421 4210%) 5.4008
(422 4220%) 6.6378
(423 4230%) 6.1315
(424 4240%) 6.6388
(425 4250%) 4.9089
(426 4260%) 6.4351
(427 4270%) 5.1482
(428 4280%) 3.0260
(429 4290%) 4.9495
(430 4300%) 3.7016
(431 4310%) 6.1299
(432 4320%) 4.0895
(433 4330%) 3.9902
(434 4340%) 5.9496
(435 4350%) 5.3130
(436 4360%) 4.6522
(437 4370%) 5.4032
(438 4380%) 4.4384
(439 4390%) 5.7474
(440 4400%) 4.3345
(441 4410%) 2.7817
(442 4420%) 4.0884
(443 4430%) 3.5009
(444 4440%) 2.8396
(445 4450%) 4.5937
(446 4460%) 3.1118
(447 4470%) 3.3859
(448 4480%) 2.5290
(449 4490%) 1.8920
(450 4500%) 5.7369
(451 4510%) 1.6092
(452 4520%) 6.4466
(453 4530%) 7.0454
(454 4540%) 5.4016
(455 4550%) 6.4004
(456 4560%) 5.4915
(457 4570%) 4.9020
(458 4580%) 2.3821
(459 4590%) 6.1554
(460 4600%) 6.0630
(461 4610%) 6.0515
(462 4620%) 2.9704

(463 4630%) 4.3882
(464 4640%) 2.5619
(465 4650%) 3.1845
(466 4660%) 3.5118
(467 4670%) 3.7234
(468 4680%) 4.5972
(469 4690%) 4.2440
(470 4700%) 5.6884
(471 4710%) 3.2573
(472 4720%) 4.5277
(473 4730%) 4.7703
(474 4740%) 3.5489
(475 4750%) 1.6232
(476 4760%) 4.6424
(477 4770%) 5.2152
(478 4780%) 2.1379
(479 4790%) 4.4938
(480 4800%) 3.1150
(481 4810%) 4.2398
(482 4820%) 3.1490
(483 4830%) 2.7097
(484 4840%) 2.2471
(485 4850%) 3.2650
(486 4860%) 2.9194
(487 4870%) 4.7739
(488 4880%) 2.2859
(489 4890%) 4.3408
(490 4900%) 4.5840
(491 4910%) 5.3370
(492 4920%) 4.5242
(493 4930%) 4.1507
(494 4940%) 5.4935
(495 4950%) 3.4002
(496 4960%) 3.6655
(497 4970%) 2.0248
(498 4980%) 3.8510
(499 4990%) 4.2003
(500 5000%) 1.5793
(501 5010%) 2.1432
(502 5020%) 1.8643
(503 5030%) 2.7311

(504 5040%) 4.2268
(505 5050%) 2.7305
(506 5060%) 3.2924
(507 5070%) 3.4501
(508 5080%) 2.6632
(509 5090%) 1.7728
(510 5100%) 2.4792
(511 5110%) 4.5720
(512 5120%) 2.8095
(513 5130%) 3.0489
(514 5140%) 6.3393
(515 5150%) 2.4366
(516 5160%) 2.4764
(517 5170%) 5.1370
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