```
In [2]:
        1 !nvidia-smi
        Mon Feb 1 16:50:04 2021
         NVIDIA-SMI 460.32.03 Driver Version: 460.32.03 CUDA Version: 11.2
                        Persistence-M | Bus-Id | Disp.A | Volatile Uncorr. ECC
         Fan Temp Perf Pwr:Usage/Cap
                                     Memory-Usage | GPU-Util Compute M.
         _____+
           0 GeForce GTX 108... Off | 00000000:18:00.0 Off |
         70% 80C P2 99W / 250W | 10809MiB / 11178MiB |
                                                            100%
                                                                     Default
           1 GeForce GTX 108... Off | 00000000:3B:00.0 Off |
                                                                     N/A
          0% 47C P8 10W / 250W | 2453MiB / 11178MiB |
                                                                    Default
                                                                        N/A
           2 GeForce GTX 108... Off | 00000000:5E:00.0 Off |
                                                                        N/A
          0% 39C P8 9W / 250W | 5887MiB / 11178MiB |
                                                                     Default
           3 GeForce GTX 108... Off | 00000000:86:00.0 Off |
                                                                        N/A
         68% 80C P2 249W / 250W | 10809MiB / 11178MiB |
                                                                     Default
                                                                        N/A
         Processes:
                                                                  GPU Memory
                            PID Type Process name
          GPU GI CI
               ID ID
                                                                  Usage
         ______
                                  G /usr/lib/xorg/Xorg
               N/A N/A
                           1730
                                                                    10801MiB
            0
               N/A N/A
                          905721
                                 C python
               N/A N/A
                          1730
                                 G /usr/lib/xorg/Xorg
                                                                       4MiB
            1
                         577506
                                 C ...khov/anaconda3/bin/python
                                                                      991MiB
               N/A N/A
               N/A N/A
                        1449170
                                 C ...a3/envs/common/bin/python
                                                                      713MiB
            1
               N/A N/A
            1
                         2155081
                                  C ...khov/anaconda3/bin/python
                                                                      741MiB
                                  G /usr/lib/xorg/Xorg
            2
               N/A N/A
                           1730
                                                                       4MiB
               N/A N/A
                                  C ...a3/envs/common/bin/python
                                                                      711MiB
            2
                          364307
               N/A N/A
                          956198
                                    C ...a3/envs/common/bin/python
                                                                      669MiB
            2
               N/A N/A
                         1751671
                                  C ...a3/envs/common/bin/python
                                                                     1471MiB
            2
               N/A N/A
                         2813905
                                  C ...a3/envs/common/bin/python
                                                                     2227MiB
                                                                      801MiB
            2 N/A N/A
                         3154363
                                  C ...a3/envs/common/bin/python
            3 N/A N/A
                           1730
                                    G /usr/lib/xorg/Xorg
                                                                       4MiB
            3 N/A N/A 1982195
                                    C python
                                                                    10801MiB
In [3]: | 1 # from models import Encoder, Decoder, Img2Caption
         2 from dataset import CocoDataset
In [3]: | 1 | import torchvision
         2 import torch
         3 import torch.nn as nn
         4 | import torch.nn.functional as F
         5 from torch.utils.data import DataLoader
         7 import numpy as np
         9 from pycocotools.coco import COCO
        10 | from nltk.translate.bleu_score import sentence_bleu
        11
        12 import tqdm
        13 | import matplotlib.pyplot as plt
        14 from IPython.display import clear_output
        16 from transformers import DebertaTokenizer, DebertaModel
In [4]: | 1 | import os
         2 os.environ["CUDA_VISIBLE_DEVICES"]="1"
         3 torch.cuda.empty_cache()
In [6]: | 1 | # training images
         2 len(os.listdir("images_train/"))
Out[6]: 14117
In [7]: | 1 | # val images
         2 len(os.listdir("images_val/"))
Out[7]: 8376
         1 # current dir
         2 os.listdir()
Out[6]: ['.idea',
         '.ipynb_checkpoints',
         'annotations',
         'best_val_model.pt',
         'dataset.py',
         'dump_1.pt',
         'images_train',
         'images_val',
         'kot.jpg',
         'models.py',
         'Training.ipynb',
         'Usage example.ipynb',
         '__pycache__']
In [8]: | 1 | # coco = COCO(annotation_file="annotations/captions_train2014.json")
         2 # coco.download("images_train", imgIds=[coco.anns[key]['image_id'] for key in coco.anns])
In [9]: | 1 # coco = COCO(annotation_file="annotations/captions_val2014.json")
         2 # coco.download("images_val", imgIds=[coco.anns[key]['image_id'] for key in coco.anns])
In [10]:
        1 tokenizer = DebertaTokenizer.from_pretrained("microsoft/deberta-base")
```

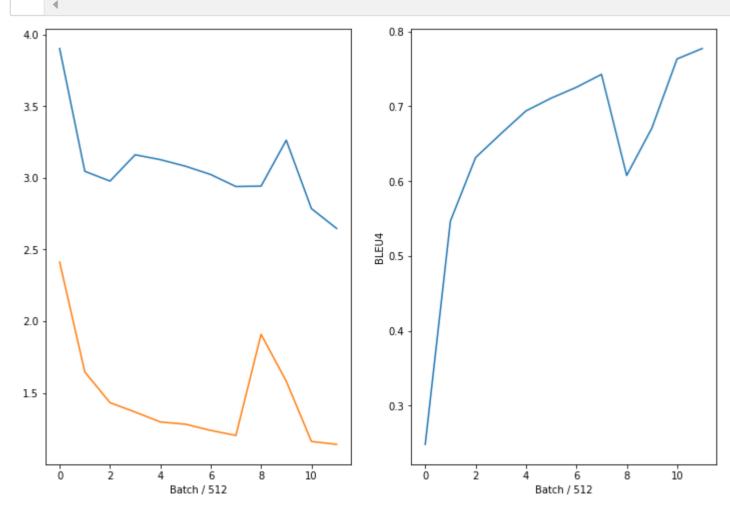
```
In [12]:
          1 def collate_fn(data):
                 images, captions = zip(*data)
           4
                 images = torch.stack(images, 0)
                 max_len = max([len(caption) for caption in captions])
                 targets = torch.zeros(len(captions), max_len).long()
                 for i, caption in enumerate(captions):
                     for j in range(len(caption)):
           9
                         targets[i, j] = caption[j]
          10
          11
                  return images.to(device), targets.to(device)
In [13]: | 1 | batch_size = 16
           train_data = CocoDataset("images_train/", "annotations/captions_train2014.json", tokenizer)
           3 test_data = CocoDataset("images_val/", "annotations/captions_val2014.json", tokenizer)
             train_loader = DataLoader(train_data,
                                       batch_size=batch_size,
                                        collate_fn=collate_fn)
           8 test_loader = DataLoader(test_data,
                                       batch_size=batch_size,
          10
                                        collate_fn=collate_fn)
         loading annotations into memory...
         Done (t=0.82s)
         creating index...
         index created!
         loading annotations into memory...
         Done (t=0.37s)
         creating index...
         index created!
In [14]:
          1 class Decoder(nn.Module):
                 def __init__(self, vocab_size, embedding_dim, encoder_dim, attention_dim, hidden_size, dropout):
                     super().__init__()
                     self.encoder_dim = encoder_dim
           5
                     self.embedding_dim = embedding_dim
           6
                     self.hidden_size = hidden_size
           8
                     self.attention_dim = attention_dim
           9
                     self.vocab_size = vocab_size
          10
          11
                     self.attention = nn.Linear(encoder_dim+embedding_dim, attention_dim)
          12
          13
                     self.rnn = nn.GRU(attention_dim, hidden_size)
          14
          15
                     self.out = nn.Linear(hidden_size+attention_dim, vocab_size)
          16
          17
                     self.dropout = nn.Dropout(dropout)
          18
          19
                  def forward(self, inputs, hidden, encoder_outputs):
          20
                      concated_inputs = torch.cat((inputs, encoder_outputs.unsqueeze(1)), dim=2)
          21
                      attented = self.dropout(torch.tanh(self.attention(concated_inputs))).permute(1, 0, 2)
          22
          23
                     output, hidden = self.rnn(attented, hidden)
          24
                      concated_outputs = F.relu(torch.cat((output, attented), dim=2))
          25
                     out = self.out(self.dropout(concated_outputs))
          26
          27
                     return out, hidden
          28
          29
                  def initHidden(self, bs):
          30
                      return torch.zeros(1, bs, self.hidden_size, device=device)
In [15]:
          1 class Img2Caption(nn.Module):
                 def __init__(self, encoder, decoder, gptModel):
                    super().__init__()
           4
           5
                      self.encoder = encoder
                      self.decoder = decoder
           7
                     self.gpt = gptModel
           8
          9
                  def forward(self, image, caption, teacher_forcing_ratio=0.5):
                       caption: bs x max_len
          10 #
          11
                      batch_size = caption.shape[0]
          12
                      max_len = caption.shape[1]
          13
                      trg_vocab_size = self.decoder.vocab_size
          14
                      outputs = torch.zeros(max_len, batch_size, trg_vocab_size)
                      encoder_outputs = self.encoder(image)
          15
          16
                      caption = gptModel(caption).last_hidden_state
          17
          18
                      first_input = caption[:, 0].unsqueeze(1)
                      hidden = decoder.initHidden(first_input.shape[0])
          19
          20
          21
                      for t in range(1, max_len):
          22
                         output, hidden = self.decoder(first_input, hidden, encoder_outputs)
          23
                          outputs[t] = output
          24
          25
                         teacher_force = np.random.random() < teacher_forcing_ratio</pre>
                         if teacher_force:
          26
          27
                             first_input = caption[:, t].unsqueeze(1)
          28
                         else:
          29
                             first_input = self.gpt(torch.argmax(output, dim=2)).last_hidden_state.permute(1, 0, 2)
          30
          31
                      return outputs
In [16]:
          1 encoder_dim = 512
           2 model_conv=torchvision.models.resnet101(pretrained=True)
           model_conv.fc = nn.Linear(in_features=2048, out_features=encoder_dim)
           5 for i, pair in enumerate(model_conv.named_children()):
                 _, child = pair
                 if len(list((model_conv.named_children()))) - i > 3:
           8
                     for _, params in child.named_parameters():
                         params.requires_grad = False
          10 | model_conv = model_conv.to(device)
In [17]: | 1 | # vocab_size, embedding_dim, encoder_dim, attention_dim, hidden_size, decoder_dim, dropout
           2 decoder = Decoder(tokenizer.vocab_size, 768, encoder_dim, 256, 256, 0.2).to(device)
             gptModel = DebertaModel.from_pretrained('microsoft/deberta-base').to(device)
           5 model = Img2Caption(model_conv, decoder, gptModel).to(device)
           7 criterion = nn.CrossEntropyLoss()
           8 optimizer = torch.optim.Adam(model.parameters(), lr=0.0001)
         Some weights of the model checkpoint at microsoft/deberta-base were not used when initializing DebertaModel: ['deberta.embeddings.position_embeddings.weight']
         - This IS expected if you are initializing DebertaModel from the checkpoint of a model trained on another task or with another architecture (e.g. initializing a BertForSequenceClas
```

sification model from a BertForPreTraining model). - This IS NOT expected if you are initializing DebertaModel from the checkpoint of a model that you expect to be exactly identical (initializing a BertForSequenceClassification mod

el from a BertForSequenceClassification model).

```
In [18]:
          1 def decode_captions(captions):
                 value = tokenizer.decode(captions)
                 print(value)
                 return value
           4
           6 def calculate_bleu(captions, targets):
                 captions = torch.argmax(captions.permute(1, 0, 2), axis=2)
                 cum_bleu = 0
          9
                 for i in range(len(captions)):
          10
                      cum_bleu += sentence_bleu(
          11
                     [decode_captions(targets[i]).split()],
          12
                      decode_captions(captions[i]).split()
          13
                     )
          14
                 print(cum_bleu)
                 return cum_bleu / len(captions)
          15
```

```
In [ ]: 1 | %matplotlib inline
          2 model.train()
          3 torch.cuda.empty_cache()
          5 logging_freq = 512
          8 max_loss = np.inf
         10 train_loss_accum = 0
         11 history = []
         12 val_bleu_history = []
         13 val_loss_history = []
         14 for epoch in range(20):
         15
                for i, batch in enumerate(tqdm.notebook.tqdm(train_loader)):
         16
         17
                    images, captions = batch
         18
         19
                     optimizer.zero_grad()
         20
                     output = model(images, captions, teacher_forcing_ratio=1/np.log(2+epoch))
         21
                    loss = criterion(output.permute(1, 0, 2).reshape(-1, tokenizer.vocab_size).cpu(), captions.view(-1).cpu())
         22
                    loss.backward()
         23
                    torch.nn.utils.clip_grad_norm_(model.parameters(), 1)
         24
         25
                     optimizer.step()
         26
                    train_loss_accum += loss
         27
                     if (i+1)%logging_freq==0:
         28
                        history.append(train_loss_accum / logging_freq)
         29
                        train_loss_accum = 0
         30
         31
                        loss_accum = 0
         32
                        bleu_accum = 0
         33
                        for test_batch in tqdm.notebook.tqdm(test_loader):
         34
                            images, captions = test_batch
         35
                            output = model(images, captions, teacher_forcing_ratio=1)
         36
                            test_loss = criterion(output.permute(1, 0, 2).reshape(-1, tokenizer.vocab_size).cpu(), captions.view(-1).cpu())
         37
                            loss_accum += test_loss.cpu().data.numpy()
         38
                            bleu_accum += calculate_bleu(output, captions)
         39
                        val_bleu_history.append(bleu_accum / len(test_loader))
         40
                        val_loss_history.append(loss_accum / len(test_loader))
         41
         42
                         clear_output(True)
         43
                        fig, ax = plt.subplots(nrows=1, ncols=2, figsize=(12, 8))
         44
         45
         46
                        ax[0].plot(history, label='train loss')
         47
                        if val_loss_history is not None:
         48
                            ax[0].plot(val_loss_history, label='validation loss')
                            ax[0].set_xlabel(f'Batch / {logging_freq}')
         49
         50
                        if val_bleu_history is not None:
                            ax[1].plot(val_bleu_history, label='validation bleu history')
         52
53
                            ax[1].set_xlabel(f'Batch / {logging_freq}')
                            ax[1].set_ylabel('BLEU4')
         54 #
                          plt.legend()
         55
                         plt.show()
         56
         57
                        if val_loss_history[-1] < max_loss:</pre>
         58
         59
                            max_loss = val_loss_history[-1]
                            torch.save(model.state_dict(), 'best_val_model.pt')
         60
```



```
2 calculate_bleu(output, captions)
A black stove sitting between counters in a kitchen.
A kitchen stove with a counters. a.
A clean kitchen with a bar and large window.
A kitchen kitchen with a bar. a..
A tile bathroom with a toilet, cabinet, sink, mirror, tub, shower curtain, and towel is shown.
A bathroom bathroom with a toilet a a and sink. mirror, and is.
a white kitchen a table and a white refrigerator
a kitchen kitchen a a and a a.
A man is chopping green peppers with a knife.
A man is of a chairs. a.
A bike is parked on a bridge without a person.
A man is a a a a.. person.
A large kitchen with a tiled floor and two refrigerators.
A kitchen kitchen with a t a a a two. kitchen.
A dog is standing and looking through the threshold into a room with sinks and and a mirror
A bathroom is a a looking in a a into a room. and
A man stands in the bathroom in front of the sink.
A kitchen stands with a bathroom with a the.
A bicycle lying on a beach with the sunset in the background.
A man lying on a beach on a in a background.
An arrangement of utensils sitting outside on a table.
An kitchen of a withils in a a a.
A newly modeled kitchen with a bicycle poster.
A kitchen bathroom with a a kitchen. a
Several people are spending time in a park with a jet airliner on the ground in the background
Several a are a a in a a on a. a. ground.
a police man in uniform is riding a red motorcycle
a man man a a is in a a motorcycle.
a small plane on an air port run way
a man plane on a air. a
A bathroom with a large bathtub and a sink.
A bathroom with a bathroom bath with a a sink.
A person getting ready to board a bus with a man on a bike in front of it.
A around a a board a a. a on a in a.
A kitchen with lots of counter space next to a window.
A kitchen with a with counter a a a.
A countertop holding a chicken and a bundt cake.
A kitchentop on a a. a..
some girls are in a kitchen and one is doing dishes
some man are with in kitchen in a doing.
a woman in a long coat and her pink umbrella
a man in a a flowers on a umbrella
A bus has bicycles mounted on the front.
A man has a a on a a.
A kitchen sink next to a window and an automatic dishwasher.
A kitchen sink with a a with. a a.washer
There are several people enjoying the fresh air and sunshine in this field.
There of several a a the with a a a. this..
A granite counter top with a sink is centered in a kitchen with a stove, cabinets, microwave, and a framed vintage advertisement for bicycles on the wall in the background.
A kitchen counter kitchen kitchen a kitchen. a in a kitchen. stove. microwave and a field. for. on the wall in the kitchen.
A bathroom with a mirror and decorations on the wall.
A bathroom with a kitchen and a and a wall.
A bathroom sink beneath a very large mirror reflecting a roll of toilet paper.
A bathroom sink and a very a a a a bathroom of a a.
A large group of bicycles are in a row.
A bathroom group a a are a a..
Three people sitting on a wooden bench eating plates of food.
Three people sitting in a wooden a a. of a.
A man is asleep with a colt on a mattress.
A man is on a a a in a a.
Three men standing around a kitchen having a drink together.
Three man standing in a kitchen a a a together.
A large bathroom with tile flooring and white fixtures
A bathroom bathroom with a floor. a
```

## Out[20]: 0.03163396867631481

1.0122869976420739

In [20]: 1 # bleu on train split

## Отчет

О том что есть в этом ноутбуке:

- 1. Что-то явно случилось с валидационной выборкой почему-то на ней запредельный bleu4, а на трейне bleu4 в несколько разм меньше. Кажется, что в дальнейшем стоит удалить данные и скачать их заново.
- 2. Модель недообучена: я долго дебажил вывод модельки, а resnet101 + deberta + декодер занимают 8гб видеопамяти спустя неделю меня выгнали с кластера.
- 3. Файлы датасета и модели отделно вынесены в dataset.py/models.py (здесь оставлены для удобства проверки).

О том чего нет в ноутбуке:

- 1. Изначально пробовал bpe токенизацию из youtokentome, не особо получилось
- 2. Пробовал w2v/обычный берт эмбединги была бага в препроцессинге текста, поэтому с ними результат не зафиксировал
- 3. Пробовал прикрутить "умный" attention с разных слоев resnet не очень получилось
- 4. Пример капшининга одной функцией в ноутбуке Usage example