

model_131_beta=5_using_the_alter_algo (1)

September 26, 2023

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[1]: # Author: William Chuang
# Last modified: Sep 26, 2023
# This notebook is built on the code written by Dr. Phillip Lippe.
#
## Standard libraries
import os
import numpy as np
import random
import math
import json
from functools import partial
import statistics as stat

## PyTorch
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.utils.data as data
import torch.optim as optim

# PyTorch Lightning
try:
    import pytorch_lightning as pl
except ModuleNotFoundError: # Google Colab does not have PyTorch Lightning
    ↪ installed by default. Hence, we do it here if necessary
    !pip install --quiet pytorch-lightning>=1.4
    import pytorch_lightning as pl
from pytorch_lightning.callbacks import LearningRateMonitor, ModelCheckpoint

# Path to the folder where the datasets are/should be downloaded (e.g. CIFAR10)
DATASET_PATH = "./data"
# Path to the folder where the pretrained models are saved
CHECKPOINT_PATH = "./saved_models/tutorial6"

# Setting the seed
pl.seed_everything(42)
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# Ensure that all operations are deterministic on GPU (if used) for
↳ reproducibility
torch.backends.cudnn.deterministic = True
torch.backends.cudnn.benchmark = False

device = torch.device("cuda:0") if torch.cuda.is_available() else torch.
↳ device("cpu")
print("Device:", device)

def scaled_dot_product(q, k, v, mask=None):
    d_k = q.size()[-1]
    PATH = "./tmp.pth"

    #torch.save(reverse_model.state_dict(), PATH)
    #w=torch.load(PATH)
    #d=sigma=torch.std((w["transformer.layers.0.self_attn.qkv_proj.weight"])).
    ↳ item()
    #print(d_k)
    attn_logits = torch.matmul(q, k.transpose(-2, -1))
    attn_logits = attn_logits * 5 #/ (math.sqrt(d_k)) #math.sqrt(0.005*d_k) #0.
    ↳ 005 #10 #3 #1.414 #(0.00001*d_k) #(d_k)**(1/100) #math.sqrt(d_k*2)
    #print(attn_logits)
    if mask is not None:
        attn_logits = attn_logits.masked_fill(mask == 0, -9e15)
    attention = F.softmax(attn_logits, dim=-1)
    values = torch.matmul(attention, v)
    return values, attention

class MultiheadAttention(nn.Module):

    def __init__(self, input_dim, embed_dim, num_heads):
        super().__init__()
        assert embed_dim % num_heads == 0, "Embedding dimension must be 0
↳ modulo number of heads."

        self.embed_dim = embed_dim
        self.num_heads = num_heads
        self.head_dim = embed_dim // num_heads

        # Stack all weight matrices 1...h together for efficiency
        # Note that in many implementations you see "bias=False" which is
↳ optional
        self.qkv_proj = nn.Linear(input_dim, 3*embed_dim)
        self.o_proj = nn.Linear(embed_dim, embed_dim)

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self._reset_parameters()

def _reset_parameters(self):
    # Original Transformer initialization, see PyTorch documentation
    nn.init.xavier_uniform_(self.qkv_proj.weight)
    self.qkv_proj.bias.data.fill_(0)
    nn.init.xavier_uniform_(self.o_proj.weight)
    self.o_proj.bias.data.fill_(0)

def forward(self, x, mask=None, return_attention=False):
    batch_size, seq_length, _ = x.size()
    if mask is not None:
        mask = expand_mask(mask)
    qkv = self.qkv_proj(x)

    # Separate Q, K, V from linear output
    qkv = qkv.reshape(batch_size, seq_length, self.num_heads, 3*self.
↪head_dim)
    qkv = qkv.permute(0, 2, 1, 3) # [Batch, Head, SeqLen, Dims]
    q, k, v = qkv.chunk(3, dim=-1)

    # Determine value outputs
    values, attention = scaled_dot_product(q, k, v, mask=mask)
    values = values.permute(0, 2, 1, 3) # [Batch, SeqLen, Head, Dims]
    values = values.reshape(batch_size, seq_length, self.embed_dim)
    o = self.o_proj(values)

    if return_attention:
        return o, attention
    else:
        return o

class EncoderBlock(nn.Module):

    def __init__(self, input_dim, num_heads, dim_feedforward, dropout=0.0):
        """
        Inputs:
        input_dim - Dimensionality of the input
        num_heads - Number of heads to use in the attention block
        dim_feedforward - Dimensionality of the hidden layer in the MLP
        dropout - Dropout probability to use in the dropout layers
        """
        super().__init__()

        # Attention layer
        self.self_attn = MultiheadAttention(input_dim, input_dim, num_heads)

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        # Two-layer MLP
        self.linear_net = nn.Sequential(
            nn.Linear(input_dim, dim_feedforward),
            nn.Dropout(dropout),
            nn.ReLU(inplace=True),
            nn.Linear(dim_feedforward, input_dim)
        )

        # Layers to apply in between the main layers
        self.norm1 = nn.LayerNorm(input_dim)
        self.norm2 = nn.LayerNorm(input_dim)
        self.dropout = nn.Dropout(dropout)

    def forward(self, x, mask=None):
        # Attention part
        attn_out = self.self_attn(x, mask=mask)
        x = x + self.dropout(attn_out)
        x = self.norm1(x)

        # MLP part
        linear_out = self.linear_net(x)
        x = x + self.dropout(linear_out)
        x = self.norm2(x)

        return x

class TransformerEncoder(nn.Module):

    def __init__(self, num_layers, **block_args):
        super().__init__()
        self.layers = nn.ModuleList([EncoderBlock(**block_args) for _ in
↪range(num_layers)])

    def forward(self, x, mask=None):
        for l in self.layers:
            x = l(x, mask=mask)
        return x

    def get_attention_maps(self, x, mask=None):
        attention_maps = []
        for l in self.layers:
            _, attn_map = l.self_attn(x, mask=mask, return_attention=True)
            attention_maps.append(attn_map)
            x = l(x)
        return attention_maps

class PositionalEncoding(nn.Module):

    def __init__(self, d_model, max_len=5000):

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    """
    Inputs
        d_model - Hidden dimensionality of the input.
        max_len - Maximum length of a sequence to expect.
    """
    super().__init__()

    # Create matrix of [SeqLen, HiddenDim] representing the positional
    ↪ encoding for max_len inputs
    pe = torch.zeros(max_len, d_model)
    position = torch.arange(0, max_len, dtype=torch.float).unsqueeze(1)
    div_term = torch.exp(torch.arange(0, d_model, 2).float() * (-math.
    ↪ log(10000.0) / d_model))
    pe[:, 0::2] = torch.sin(position * div_term)
    pe[:, 1::2] = torch.cos(position * div_term)
    pe = pe.unsqueeze(0)

    # register_buffer => Tensor which is not a parameter, but should be
    ↪ part of the modules state.
    # Used for tensors that need to be on the same device as the module.
    # persistent=False tells PyTorch to not add the buffer to the state
    ↪ dict (e.g. when we save the model)
    self.register_buffer('pe', pe, persistent=False)

    def forward(self, x):
        x = x + self.pe[:, :x.size(1)]
        return x
class TransformerPredictor(pl.LightningModule):

    def __init__(self, input_dim, model_dim, num_classes, num_heads,
    ↪ num_layers, lr, warmup, max_iters, dropout=0.0, input_dropout=0.0):
        """
        Inputs:
            input_dim - Hidden dimensionality of the input
            model_dim - Hidden dimensionality to use inside the Transformer
            num_classes - Number of classes to predict per sequence element
            num_heads - Number of heads to use in the Multi-Head Attention
    ↪ blocks
            num_layers - Number of encoder blocks to use.
            lr - Learning rate in the optimizer
            warmup - Number of warmup steps. Usually between 50 and 500
            max_iters - Number of maximum iterations the model is trained for.
    ↪ This is needed for the CosineWarmup scheduler
            dropout - Dropout to apply inside the model
            input_dropout - Dropout to apply on the input features
        """

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    super().__init__()
    self.save_hyperparameters()
    self._create_model()

    def _create_model(self):
        # Input dim -> Model dim
        self.input_net = nn.Sequential(
            nn.Dropout(self.hparams.input_dropout),
            nn.Linear(self.hparams.input_dim, self.hparams.model_dim)
        )
        # Positional encoding for sequences
        self.positional_encoding = PositionalEncoding(d_model=self.hparams.
↪model_dim)
        # Transformer
        self.transformer = TransformerEncoder(num_layers=self.hparams.
↪num_layers,
                                           input_dim=self.hparams.model_dim,
                                           dim_feedforward=2*self.hparams.
↪model_dim,
                                           num_heads=self.hparams.num_heads,
                                           dropout=self.hparams.dropout)

        # Output classifier per sequence element
        self.output_net = nn.Sequential(
            nn.Linear(self.hparams.model_dim, self.hparams.model_dim),
            nn.LayerNorm(self.hparams.model_dim),
            nn.ReLU(inplace=True),
            nn.Dropout(self.hparams.dropout),
            nn.Linear(self.hparams.model_dim, self.hparams.num_classes)
        )

    def forward(self, x, mask=None, add_positional_encoding=True):
        """
        Inputs:
            x - Input features of shape [Batch, SeqLen, input_dim]
            mask - Mask to apply on the attention outputs (optional)
            add_positional_encoding - If True, we add the positional encoding_
↪to the input.

            Might not be desired for some tasks.
        """
        x = self.input_net(x)
        if add_positional_encoding:
            x = self.positional_encoding(x)
        x = self.transformer(x, mask=mask)
        x = self.output_net(x)
        return x

    @torch.no_grad()

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def get_attention_maps(self, x, mask=None, add_positional_encoding=True):
    """
    Function for extracting the attention matrices of the whole Transformer
    for a single batch.
    Input arguments same as the forward pass.
    """
    x = self.input_net(x)
    if add_positional_encoding:
        x = self.positional_encoding(x)
    attention_maps = self.transformer.get_attention_maps(x, mask=mask)
    return attention_maps

def configure_optimizers(self):
    optimizer = optim.Adam(self.parameters(), lr=self.hparams.lr)

    # Apply lr scheduler per step
    lr_scheduler = CosineWarmupScheduler(optimizer,
                                         warmup=self.hparams.warmup,
                                         max_iters=self.hparams.max_iters)
    return [optimizer], [{'scheduler': lr_scheduler, 'interval': 'step'}]

def training_step(self, batch, batch_idx):
    raise NotImplementedError

def validation_step(self, batch, batch_idx):
    raise NotImplementedError

def test_step(self, batch, batch_idx):
    raise NotImplementedError

class ReverseDataset(data.Dataset):

    def __init__(self, num_categories, seq_len, size):
        super().__init__()
        self.num_categories = num_categories

        self.seq_len = seq_len
        self.size = size

        self.data = torch.randint(self.num_categories, size=(self.size, self.
    seq_len))
        # self.data = torch.abs(torch.normal(0, 1, size=(self.size, self.
    seq_len))).long())
        # torch.randint(low=0, high, size, \*, generator=None, out=None,
    dtype=None,
        # layout=torch.strided, device=None, requires_grad=False) → Tensor
        print(self.num_categories)
        print(self.seq_len)

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        print(self.size)
        print(self.data)

    def __len__(self):
        return self.size

    def __getitem__(self, idx):
        inp_data = self.data[idx]
        labels = torch.flip(inp_data, dims=(0,))
        return inp_data, labels
''' Examples of torch.randint
#>>> torch.randint(3, 5, (3,))
#tensor([4, 3, 4])

#>>> torch.randint(10, (2, 2))
#tensor([[0, 2],
#        [5, 5]])

#>>> torch.randint(3, 10, (2, 2))
#tensor([[4, 5],
#        [6, 7]])'''

#'''>>> torch.normal(mean=0.5, std=torch.arange(1., 6.))
#tensor([-1.2793, -1.0732, -2.0687,  5.1177, -1.2303])'''

class ReversePredictor(TransformerPredictor):

    def _calculate_loss(self, batch, mode="train"):
        # Fetch data and transform categories to one-hot vectors
        inp_data, labels = batch
        inp_data = F.one_hot(inp_data, num_classes=self.hparams.num_classes).
        ↪float()

        # Perform prediction and calculate loss and accuracy
        preds = self.forward(inp_data, add_positional_encoding=True)
        loss = F.cross_entropy(preds.view(-1, preds.size(-1)), labels.view(-1))
        acc = (preds.argmax(dim=-1) == labels).float().mean()

        # Logging
        self.log(f"{mode}_loss", loss)
        self.log(f"{mode}_acc", acc)
        return loss, acc

    def training_step(self, batch, batch_idx):
        loss, _ = self._calculate_loss(batch, mode="train")

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        return loss

    def validation_step(self, batch, batch_idx):
        _ = self._calculate_loss(batch, mode="val")

    def test_step(self, batch, batch_idx):
        _ = self._calculate_loss(batch, mode="test")
def train_reverse(**kwargs):

    # Create a PyTorch Lightning trainer with the generation callback
    root_dir = os.path.join(CHECKPOINT_PATH, "ReverseTask")
    os.makedirs(root_dir, exist_ok=True)
    trainer = pl.Trainer(default_root_dir=root_dir,
                        callbacks=[ModelCheckpoint(save_weights_only=True,
↪mode="max", monitor="val_acc")],
                        accelerator="gpu" if str(device).startswith("cuda")
↪else "cpu",
                        devices=1,
                        max_epochs=10,
                        gradient_clip_val=5)
    trainer.logger._default_hp_metric = None # Optional logging argument that
↪we don't need
    trainer.callbacks
    # Check whether pretrained model exists. If yes, load it and skip training
    pretrained_filename = os.path.join(CHECKPOINT_PATH, "ReverseTask.ckpt")
    if os.path.isfile(pretrained_filename):
        print("Found pretrained model, loading...")
        model = ReversePredictor.load_from_checkpoint(pretrained_filename)
    else:
        print("Found pretrained model does not exist, generating...")
        model = ReversePredictor(max_iters=trainer.
↪max_epochs*len(train_loader), **kwargs)
        trainer.fit(model, train_loader, val_loader)

    # Test best model on validation and test set
    val_result = trainer.test(model, val_loader, verbose=False)
    test_result = trainer.test(model, test_loader, verbose=False)
    result = {"test_acc": test_result[0]["test_acc"], "val_acc":
↪val_result[0]["test_acc"]}

    model = model.to(device)
    return model, result
class CosineWarmupScheduler(optim.lr_scheduler._LRScheduler):

    def __init__(self, optimizer, warmup, max_iters):
        self.warmup = warmup
        self.max_num_iters = max_iters

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        super().__init__(optimizer)

    def get_lr(self):
        lr_factor = self.get_lr_factor(epoch=self.last_epoch)
        return [base_lr * lr_factor for base_lr in self.base_lrs]

    def get_lr_factor(self, epoch):
        lr_factor = 0.5 * (1 + np.cos(np.pi * epoch / self.max_num_iters))
        if epoch <= self.warmup:
            lr_factor *= epoch * 1.0 / self.warmup
        return lr_factor

dataset = partial(ReverseDataset, 100, 20)
train_loader = data.DataLoader(dataset(15000), batch_size=128, shuffle=True,
    ↪drop_last=True, pin_memory=True)
val_loader = data.DataLoader(dataset(1000), batch_size=128)
test_loader = data.DataLoader(dataset(100000), batch_size=128)
inp_data, labels = train_loader.dataset[0]
print("Input data:", inp_data)
print("Labels: ", labels)

```

INFO:lightning_fabric.utilities.seed:Global seed set to 42

Device: cuda:0

100

20

15000

```

tensor([[42, 67, 76, ..., 95, 67, 6],
        [49, 76, 73, ..., 76, 32, 10],
        [86, 22, 77, ..., 84, 78, 8],
        ...,
        [ 1, 31, 80, ..., 38, 66, 80],
        [37, 57, 93, ..., 31, 8, 65],
        [23, 88, 33, ..., 57, 23, 51]])

```

100

20

1000

```

tensor([[ 6, 25, 53, ..., 40, 80, 91],
        [30, 12, 95, ..., 42, 22, 95],
        [50, 62, 71, ..., 39, 51, 82],
        ...,
        [45, 54, 2, ..., 76, 24, 92],
        [ 1, 50, 81, ..., 3, 23, 81],
        [93, 89, 53, ..., 37, 78, 83]])

```

100

20

100000

```

tensor([[83, 59, 87, ..., 50, 18, 19],

```

```

[90, 85, 37, ..., 31, 49, 62],
[22, 2, 57, ..., 91, 44, 21],
...,
[14, 9, 53, ..., 34, 61, 49],
[20, 83, 70, ..., 61, 43, 64],
[59, 97, 69, ..., 28, 75, 83]])
Input data: tensor([42, 67, 76, 14, 26, 35, 20, 24, 50, 13, 78, 14, 10, 54, 31,
72, 15, 95,
67, 6])
Labels: tensor([ 6, 67, 95, 15, 72, 31, 54, 10, 14, 78, 13, 50, 24, 20, 35,
26, 14, 76,
67, 42])

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[2]: reverse_model, reverse_result = train_reverse(input_dim=train_loader.dataset.
    ↪ num_categories,
                                model_dim=20,
                                num_heads=1,
                                num_classes=train_loader.dataset.
    ↪ num_categories,
                                num_layers=1,
                                dropout=0.0,
                                lr=5e-4,
                                warmup=50)

```

INFO:pytorch_lightning.utilities.rank_zero:GPU available: True (cuda), used: True

INFO:pytorch_lightning.utilities.rank_zero:TPU available: False, using: 0 TPU cores

INFO:pytorch_lightning.utilities.rank_zero:IPU available: False, using: 0 IPUs

INFO:pytorch_lightning.utilities.rank_zero:HPU available: False, using: 0 HPUs

Found pretrained model does not exist, generating...

WARNING:pytorch_lightning.loggers.tensorboard:Missing logger folder:

saved_models/tutorial6/ReverseTask/lightning_logs

INFO:pytorch_lightning.accelerators.cuda:LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES: [0]

INFO:pytorch_lightning.callbacks.model_summary:

	Name	Type	Params
0	input_net	Sequential	2.0 K
1	positional_encoding	PositionalEncoding	0
2	transformer	TransformerEncoder	3.4 K
3	output_net	Sequential	2.6 K
8.0 K	Trainable params		
0	Non-trainable params		
8.0 K	Total params		
0.032	Total estimated model params size (MB)		

Sanity Checking: 0it [00:00, ?it/s]

Training: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

Validation: 0it [00:00, ?it/s]

INFO:pytorch_lightning.utilities.rank_zero:`Trainer.fit` stopped:
`max_epochs=10` reached.

INFO:pytorch_lightning.accelerators.cuda:LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES:
[0]

Testing: 0it [00:00, ?it/s]

INFO:pytorch_lightning.accelerators.cuda:LOCAL_RANK: 0 - CUDA_VISIBLE_DEVICES:
[0]

Testing: 0it [00:00, ?it/s]

```
[3]: # @title # model_131_beta=5_using_the_alter_algo.ipynb
print(f"Val accuracy: {(100.0 * reverse_result['val_acc']):4.2f}%")
print(f"Test accuracy: {(100.0 * reverse_result['test_acc']):4.2f}%")
```

Val accuracy: 96.43%

Test accuracy: 96.44%

```
[6]: PATH = "./new.pth"
torch.save(reverse_model.state_dict(), PATH)
PATH = "./new.pth"
w=torch.load(PATH)
```

```
[7]: #for odict in w:
#     print(odict)
for odict in w:
    print(str(odict)+" : "+str(len(w[str(odict)])))
    #print(str( (w[str(odict)])[5])))
```

input_net.1.weight: 20

input_net.1.bias: 20

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transformer.layers.0.self_attn.qkv_proj.weight: 60
transformer.layers.0.self_attn.qkv_proj.bias: 60
transformer.layers.0.self_attn.o_proj.weight: 20
transformer.layers.0.self_attn.o_proj.bias: 20
transformer.layers.0.linear_net.0.weight: 40
transformer.layers.0.linear_net.0.bias: 40
transformer.layers.0.linear_net.3.weight: 20
transformer.layers.0.linear_net.3.bias: 20
transformer.layers.0.norm1.weight: 20
transformer.layers.0.norm1.bias: 20
transformer.layers.0.norm2.weight: 20
transformer.layers.0.norm2.bias: 20
output_net.0.weight: 20
output_net.0.bias: 20
output_net.1.weight: 20
output_net.1.bias: 20
output_net.4.weight: 100
output_net.4.bias: 100

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[8]: for odict in w:
      print(odict)
      print(w[str(odict)])
      #print(str( (w[str(odict)][5])))
      ↵
      ↪print("-----")

```

```

input_net.1.weight
tensor([[[-0.0350,  0.0685, -0.0798, ...,  0.0073,  0.0280, -0.0734],
         [-0.0593, -0.0045, -0.0622, ...,  0.0790, -0.0287,  0.0446],
         [-0.0612, -0.0769,  0.0088, ...,  0.0126,  0.1124, -0.0317],
         ...,
         [-0.0486, -0.0088,  0.0556, ...,  0.0201, -0.0175,  0.0331],
         [ 0.0679,  0.1667,  0.0706, ..., -0.0746,  0.0204, -0.0261],
         [-0.0220, -0.0633, -0.0132, ..., -0.0077,  0.0529,  0.0316]],
        device='cuda:0')

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input_net.1.bias
tensor([-0.0078, -0.0101, -0.0319, -0.0678, -0.0408, -0.0384, -0.0641, -0.0252,
        0.0296,  0.0166, -0.0713, -0.1129,  0.0168, -0.0400,  0.0400, -0.1022,
        0.0672, -0.0757,  0.1145, -0.0316], device='cuda:0')

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transformer.layers.0.self_attn.qkv_proj.weight
tensor([[[-0.3153, -0.2766, -0.1042, ...,  0.2538, -0.0402, -0.1034],
         [ 0.4574, -0.4184, -0.0770, ..., -0.1444,  0.1758, -0.1419],
         [-0.2668,  0.1546, -0.2720, ..., -0.1659,  0.0114, -0.1847],
         ...,
         [ 0.0966,  0.0813, -0.2479, ...,  0.1152,  0.1799, -0.1152],
         [ 0.1382,  0.0292, -0.0520, ..., -0.2155,  0.3142,  0.0095],

```

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[ 0.1878,  0.0312, -0.2110, ..., -0.2603,  0.4794,  0.0636]],
device='cuda:0')
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-----
transformer.layers.0.self_attn.qkv_proj.bias
```

```
tensor([-1.7811e-02,  2.9813e-03, -2.3117e-02,  2.2288e-03, -1.7551e-02,
        2.6426e-02,  7.9288e-03, -1.1374e-02,  1.1349e-02, -3.3630e-02,
        3.2969e-02,  3.9939e-02,  1.1011e-02, -3.2553e-02,  1.7364e-04,
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transformer.layers.0.self_attn.o_proj.weight
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-1.1767e-01, 2.7378e-02, -3.1754e-02, 3.4474e-02, -1.4763e-01,
-3.9290e-02, -1.6423e-01, 2.1083e-01, -9.4170e-02, 1.2826e-02]],
device='cuda:0')

```

```

-----
transformer.layers.0.linear_net.3.bias
tensor([ 0.1772,  0.1042, -0.0070, -0.0606, -0.1253,  0.0093,  0.1087, -0.1118,
         0.0570, -0.0111,  0.1154, -0.0812, -0.1314, -0.0398, -0.0337,  0.1071,
         0.0260, -0.1680, -0.1201,  0.1325], device='cuda:0')

```

```

-----
transformer.layers.0.norm1.weight
tensor([0.8378, 0.9346, 0.9655, 0.9374, 0.9938, 0.9402, 1.1050, 1.0993, 1.1619,
        1.0821, 1.0516, 0.9468, 0.9618, 1.0136, 1.0113, 1.0311, 0.9660, 1.0284,
        1.0776, 1.0489], device='cuda:0')
-----

```

```
transformer.layers.0.norm1.bias
tensor([ 0.0292,  0.0089,  0.0222,  0.0080, -0.0016,  0.0029,  0.0272, -0.0164,
         0.0020,  0.0066, -0.0232, -0.0488,  0.0105, -0.0039, -0.0296, -0.0061,
        -0.0089, -0.0180,  0.0039,  0.0197], device='cuda:0')
```

```
transformer.layers.0.norm2.weight
tensor([0.8659, 0.9843, 0.9372, 0.8484, 1.0230, 0.9267, 1.0689, 1.0865, 1.1448,
        1.0711, 1.0655, 0.8714, 1.0384, 0.9916, 1.0004, 1.1651, 0.9808, 1.0937,
        1.0077, 1.0053], device='cuda:0')
```

```
transformer.layers.0.norm2.bias
tensor([ 0.0124,  0.0734, -0.0151,  0.0167, -0.0241, -0.0209,  0.0060,  0.0014,
        -0.0301, -0.0304, -0.0007, -0.0601,  0.0095, -0.0379,  0.0556, -0.0056,
        -0.0231, -0.0298,  0.0462,  0.0068], device='cuda:0')
```

```
output_net.0.weight
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         -2.3525e-01, -1.8338e-01,  2.3969e-01, -2.3821e-01,  1.0900e-01,
          2.3849e-02,  4.4907e-02,  1.8865e-01,  8.0013e-02,  1.1968e-01],
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         -1.7788e-02, -6.9832e-02,  1.1112e-01, -7.1721e-02, -1.4257e-01,
         -1.1865e-01,  2.9975e-02, -1.0295e-01, -1.4501e-01,  4.3870e-02,
          5.9104e-02, -1.2452e-01,  8.7396e-03, -9.1330e-02, -2.2758e-03],
        [ 2.7073e-02,  1.6152e-01, -7.3733e-02, -2.9963e-02,  6.3972e-02,
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```

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```



```

device='cuda:0')
-----
output_net.0.bias
tensor([ 2.5755e-01,  1.3733e-01,  3.1281e-02,  1.1338e-01,  7.9566e-02,
        -2.2724e-02, -1.3715e-01, -4.1091e-02, -1.4288e-01, -3.9885e-02,
         7.9907e-02, -1.5303e-01, -1.4051e-01,  1.3282e-01,  1.6469e-01,
        -4.2821e-02,  2.4868e-02, -8.7467e-02,  1.0265e-02,  8.6914e-05],
        device='cuda:0')
-----
output_net.1.weight
tensor([1.3395, 1.2216, 1.3000, 1.2510, 0.9916, 1.2529, 1.2541, 1.2904, 1.2786,
        1.2774, 1.2715, 1.2821, 1.2662, 1.2107, 1.2897, 1.3354, 1.2779, 1.3821,
        1.2726, 1.2521], device='cuda:0')
-----
output_net.1.bias
tensor([0.3203, 0.2338, 0.2468, 0.2269, 0.0113, 0.2341, 0.2318, 0.2501, 0.2379,
        0.2432, 0.2446, 0.2684, 0.2427, 0.2111, 0.2420, 0.3019, 0.2496, 0.3022,
        0.2384, 0.2296], device='cuda:0')
-----
output_net.4.weight
tensor([[[-0.2682, -0.2782, -0.0398, ..., -0.3410,  0.3174, -0.3753],
         [ 0.4023,  0.2012, -0.3187, ..., -0.4614, -0.4321, -0.1043],
         [ 0.1181, -0.3251, -0.2473, ..., -0.3905, -0.1780,  0.3544],
         ...,
         [-0.0820, -0.1033,  0.0224, ...,  0.3858, -0.2941, -0.2559],
         [-0.1543, -0.0640,  0.1649, ..., -0.3901, -0.3766, -0.2520],
         [-0.4253, -0.0671,  0.0097, ...,  0.4040,  0.5006,  0.0212]],
        device='cuda:0')
-----
output_net.4.bias
tensor([-0.0247,  0.1526,  0.0850, -0.1464,  0.1352,  0.0385,  0.0652, -0.1488,
         0.1002,  0.1606, -0.0224,  0.1837, -0.1837, -0.1388, -0.1809,  0.1591,
         0.0405,  0.0793, -0.2122, -0.0332,  0.3101, -0.0276, -0.0770, -0.1425,
         0.1671, -0.0681,  0.0766, -0.1864,  0.1755, -0.1156,  0.1637, -0.1354,
        -0.0619, -0.3599, -0.0642,  0.2843,  0.0645, -0.2369,  0.1238, -0.0381,
         0.0584,  0.2338, -0.0972, -0.1162,  0.0536,  0.0171,  0.2094,  0.0089,
         0.0999,  0.0712, -0.1499, -0.0462, -0.0497,  0.2528, -0.0414, -0.0873,
        -0.0703, -0.0708,  0.1032,  0.0742, -0.1147, -0.0676, -0.2692, -0.0036,
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         0.1281,  0.1445,  0.0982, -0.2242], device='cuda:0')
-----

```

[]: