```
! pip install setuptools==66
! pip install d2l==1.0.0b0
!pip install matplotlib_inline
! pip install ptflops
import time
import torch
import torchvision
from torchvision
from torchvision import transforms
from d2l import torch as d2l
from ptflops import get_model_complexity_info

C→ /usr/local/lib/python3.9/dist-packages/torch/cuda/_init__.py:497: UserWarning: Can't initialize NVML
warnings.warn("Can't initialize NVML")
```

→ D2L Stuff

```
class HyperParameters:
    def save_hyperparameters(self, ignore=[]):
        raise NotImplemented
def add_to_class(Class):
    def wrapper(obj):
        setattr(Class, obj.__name__, obj)
    return wrapper
class ProgressBoard(d21.HyperParameters):
    def __init__(self, xlabel=None, ylabel=None, xlim=None,
                 ylim=None, xscale='linear', yscale='linear',
                 ls=['-', '--', '--', ':'], colors=['C0', 'C1', 'C2', 'C3'],
                 fig=None, axes=None, figsize=(3.5, 2.5), display=True):
        self.save_hyperparameters()
    def draw(self, x, y, label, every n=1):
        raise NotImplemented
class Module(nn.Module, d21.HyperParameters):
    def __init__(self, plot_train_per_epoch=2, plot_valid_per_epoch=1):
        super().__init__()
        self.save_hyperparameters()
        self.board = ProgressBoard()
    def loss(self, y_hat, y):
        raise NotImplementedError
    def forward(self, X):
        assert hasattr(self, 'net'), 'Neural network is defined'
        return self.net(X)
    def plot(self, key, value, train):
        """Plot a point in animation."""
        assert hasattr(self, 'trainer'), 'Trainer is not inited'
        self.board.xlabel = 'epoch'
        if train:
            x = self.trainer.train batch idx / \
                self.trainer.num_train_batches
            n = self.trainer.num_train_batches / \
                self.plot_train_per_epoch
            x = self.trainer.epoch + 1
            n = self.trainer.num_val_batches / \
```

```
self.plot_valid_per_epoch
        self.board.draw(x, value.to(d21.cpu()).detach().numpy(),
                        ('train_' if train else 'val_') + key,
                        every_n=int(n))
    def training_step(self, batch):
        1 = self.loss(self(*batch[:-1]), batch[-1])
        self.plot('loss', 1, train=True)
        return 1
   def validation_step(self, batch):
        1 = self.loss(self(*batch[:-1]), batch[-1])
        self.plot('loss', 1, train=False)
        return 1
    def configure_optimizers(self):
       raise NotImplementedError
class Trainer(d21.HyperParameters):
   def __init__(self, max_epochs, num_gpus=0, gradient_clip_val=0):
        self.save_hyperparameters()
        assert num_gpus == 0, 'No GPU support yet'
    def prepare_data(self, data):
        self.train_dataloader = data.train_dataloader()
        self.val_dataloader = data.val_dataloader()
        self.num_train_batches = len(self.train_dataloader)
        self.num_val_batches = (len(self.val_dataloader)
                                if self.val_dataloader is not None else 0)
    def prepare_model(self, model):
        model.trainer = self
        model.board.xlim = [0, self.max epochs]
        self.model = model
   def fit(self, model, data):
        self.prepare data(data)
        self.prepare model(model)
        self.optim = model.configure_optimizers()
        self.epoch = 0
        self.train_batch_idx = 0
        self.val batch idx = 0
        for self.epoch in range(self.max_epochs):
            self.fit_epoch()
    def fit epoch(self):
        raise NotImplementedError
@d21.add to class(d21.Trainer)
def prepare_batch(self, batch):
   return batch
@d21.add to class(d21.Trainer)
def fit_epoch(self):
   self.model.train()
   #self.totalLoss = 0
   #self.epochLoss = 0
   #self.totalValLoss = 0
   #self.valEpochLoss = 0
   #self.lossHist = []
    for batch in self.train_dataloader:
        loss = self.model.training_step(self.prepare_batch(batch))
        #self.totalLoss = self.totalLoss + loss.item()
        self.optim.zero_grad()
        with torch.no_grad():
            loss.backward()
            if self.gradient_clip_val > 0: # To be discussed later
                self.clip_gradients(self.gradient_clip_val, self.model)
            self.optim.step()
        self.train_batch_idx += 1
    #self.loaderLength = len(self.train_dataloader)
    #self.epochLoss = self.totalLoss / self.loaderLength
```

```
#print("Epoch loss: ", self.epochLoss)
   #self.lossHist.append(self.epochLoss)
   #globEpochLoss.append(self.epochLoss)
   if self.val_dataloader is None:
       return
   self.model.eval()
    for batch in self.val_dataloader:
       with torch.no_grad():
            self.model.validation_step(self.prepare_batch(batch))
        #valLoss = self.model.validation_step(self.prepare_batch(batch))
        #self.totalValLoss = self.totalValLoss + valLoss
        self.val_batch_idx += 1
   #self.valLoaderLength = len(self.val_dataloader)
   #self.valEpochLoss = self.totalValLoss / self.valLoaderLength
    #globValLoss.append(self.valEpochLoss)
class DataModule(d21.HyperParameters):
   def __init__(self, root='../data', num_workers=4):
        self.save_hyperparameters()
   def get_dataloader(self, train):
        raise NotImplementedError
   def train dataloader(self):
        return self.get_dataloader(train=True)
   def val_dataloader(self):
        return self.get_dataloader(train=False)
```

Preparing the Dataset

```
import collections
import random
import re
class TimeMachine(d21.DataModule):
    """The Time Machine dataset."""
    def _download(self):
        fname = d21.download(d21.DATA URL + 'timemachine.txt', self.root,
                                '090b5e7e70c295757f55df93cb0a180b9691891a')
        with open(fname) as f:
             return f.read()
data = TimeMachine()
raw_text = data._download()
raw_text[:60]
     Downloading ../data/timemachine.txt from <a href="http://d21-data.s3-accelerate.amazonaws.com/timemachine.txt">http://d21-data.s3-accelerate.amazonaws.com/timemachine.txt</a> ...
     'The Time Machine, by H. G. Wells [1898]\n\n\n\n\n\nThe Time Tra'
@d21.add_to_class(TimeMachine)
def _preprocess(self, text):
    return re.sub('[^A-Za-z]+', ' ', text).lower()
text = data._preprocess(raw_text)
text[:60]
     'the time machine by h g wells i the time traveller for so it'
@d21.add_to_class(TimeMachine)
def tokenize(self, text):
    return list(text)
tokens = data._tokenize(text)
','.join(tokens[:30])
```

```
't,h,e, ,t,i,m,e, ,m,a,c,h,i,n,e, ,b,y, ,h, ,g, ,w,e,l,l,s, '
  class Vocab:
      """Vocabulary for text."""
      def __init__(self, tokens=[], min_freq=0, reserved_tokens=[]):
          # Flatten a 2D list if needed
          if tokens and isinstance(tokens[0], list):
              tokens = [token for line in tokens for token in line]
          # Count token frequencies
          counter = collections.Counter(tokens)
          self.token_freqs = sorted(counter.items(), key=lambda x: x[1],
                                     reverse=True)
          # The list of unique tokens
          self.idx_to_token = list(sorted(set(['<unk>'] + reserved_tokens + [
              token for token, freq in self.token_freqs if freq >= min_freq])))
          self.token_to_idx = {token: idx
                                for idx, token in enumerate(self.idx to token)}
      def __len__(self):
          return len(self.idx_to_token)
      def __getitem__(self, tokens):
          if not isinstance(tokens, (list, tuple)):
              return self.token_to_idx.get(tokens, self.unk)
          return [self.__getitem__(token) for token in tokens]
      def to_tokens(self, indices):
          if hasattr(indices, '__len__') and len(indices) > 1:
              return [self.idx_to_token[int(index)] for index in indices]
          return self.idx_to_token[indices]
      @property
      def unk(self): # Index for the unknown token
          return self.token to idx['<unk>']
  vocab = Vocab(tokens)
  indices = vocab[tokens[:10]]
  print('indices:', indices)
  print('words:', vocab.to_tokens(indices))
       indices: [21, 9, 6, 0, 21, 10, 14, 6, 0, 14]
words: ['t', 'h', 'e', '', 't', 'i', 'm', 'e', '', 'm']
  @d21.add to class(TimeMachine)
  def build(self, raw_text, vocab=None):
      tokens = self._tokenize(self._preprocess(raw_text))
      if vocab is None: vocab = Vocab(tokens)
      corpus = [vocab[token] for token in tokens]
      return corpus, vocab
  corpus, vocab = data.build(raw_text)
  len(corpus), len(vocab)
       (173428, 28)

→ 1a) GRU

  class GRU(d21.RNN):
      def __init__(self, num_inputs, num_hiddens):
          d21.Module.__init__(self)
          self.save_hyperparameters()
          self.rnn = nn.GRU(num_inputs, num_hiddens)
  data = d21.TimeMachine(batch_size=1024, num_steps=32)
  trainer = d21.Trainer(max_epochs=50, gradient_clip_val=1, num_gpus=1)
```

Default network from example

```
gru = GRU(num_inputs=len(data.vocab), num_hiddens=32)
model = d21.RNNLM(gru, vocab_size=len(data.vocab), lr=4)
trainer.fit(model, data)
# Train time: 1min 44s
      20
                               train_ppl
                             -- val_ppl
      15
      10
       5
              10
                    20
                           30
                                 40
                                       50
                      epoch
model.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has the the the the'
model.predict('we are always', 20, data.vocab, d2l.try_gpu())
     'we are alwaysed the time the time'
model.predict('he looked across', 20, data.vocab, d21.try_gpu())
    'he looked across the time the time t'
macs, params = get_model_complexity_info(gru, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
                        flops or params are already defined for the moduleGRU ptflops can affect your code!
    Warning: variables
    Warning: module GRU is treated as a zero-op.
      5.95 k, 100.000% Params, 6.32 MMac, 100.000% MACs,
      (rnn): GRU(5.95 k, 100.000% Params, 6.32 MMac, 100.000% MACs, 28, 32)
    Computational complexity: 6.32 MMac
    Number of parameters: 5.95 k
# Halfing the number of hidden states to 16
gru16 = GRU(num_inputs=len(data.vocab), num_hiddens=16)
gruModel16 = d21.RNNLM(gru16, vocab_size=len(data.vocab), lr=4)
trainer.fit(gruModel16, data)
# Training time: 1min 31sec
      25
                               train_ppl
                              val_ppl
      20
      15
      10
```

```
gruModel16.predict('it has', 20, data.vocab, d21.try_gpu())
```

30

50

0

10

20

'it has mong the the th'

epoch

```
gruModel16.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always the the the the
gruModel16.predict('he looked across', 20, data.vocab, d21.try_gpu())
    'he looked acrossing the the the '
macs, params = get_model_complexity_info(gru16, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module GRU is treated as a zero-op.
    GRU(
      2.21 k, 100.000% Params, 2.38 MMac, 100.000% MACs,
      (rnn): GRU(2.21 k, 100.000% Params, 2.38 MMac, 100.000% MACs, 28, 16)
    Computational complexity: 2.38 MMac
    Number of parameters: 2.21 \text{ k}
# Increasing the number of hidden states to 48
gru48 = GRU(num inputs=len(data.vocab), num hiddens=48)
gruModel48 = d21.RNNLM(gru48, vocab size=len(data.vocab), lr=4)
trainer.fit(gruModel48, data)
# Training time: 2mins 14sec
                               train_ppl
     20
                               val_ppl
     15
     10
       5
                    20
        0
              10
                           30
                                 40
                                       50
                      epoch
gruModel48.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has interical mentered '
gruModel48.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always this space the psyc'
gruModel48.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked across for instance the ti'
macs, params = get_model_complexity_info(gru48, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module GRU is treated as a zero-op.
    GRU (
      11.23 k, 100.000% Params, 11.85 MMac, 100.000% MACs,
      (rnn): GRU(11.23 k, 100.000% Params, 11.85 MMac, 100.000% MACs, 28, 48)
    Computational complexity: 11.85 MMac
    Number of parameters: 11.23 k
# Increasing the number of hidden states to 64
gru64 = GRU(num_inputs=len(data.vocab), num_hiddens=64)
gruModel64 = d21.RNNLM(gru64, vocab_size=len(data.vocab), lr=4)
```

```
trainer.fit(gruModel64, data)
# Training time: 2min 43sec
```

```
train_ppl
      20
                                val ppl
      15
     10
       5
                     20
              10
                           30
                                 40
                                        50
                      epoch
gruModel64.predict('it has', 20, data.vocab, d21.try_gpu())
     'it has in thing that that '
gruModel64.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always dimension in i sume'
gruModel64.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked across the psychologist yo'
macs, params = get_model_complexity_info(gru64, (1024, 28), as_strings=True,
                                            print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
     Warning: module GRU is treated as a zero-op.
      18.05 k, 100.000% Params, 18.94 MMac, 100.000% MACs,
      (rnn): GRU(18.05 k, 100.000% Params, 18.94 MMac, 100.000% MACs, 28, 64)
     Computational complexity: 18.94 MMac
    Number of parameters: 18.05 k
# Increasing the number of hidden states to 128
gru128 = GRU(num_inputs=len(data.vocab), num_hiddens=128)
gruModel128 = d21.RNNLM(gru128, vocab_size=len(data.vocab), lr=4)
trainer.fit(gruModel128, data)
# Training time: 5min 33sec
                                train_ppl
      20
                                val ppl
      15
      10
      5
        0
              10
                     20
                           30
                                 40
                                        50
gruModel128.predict('it has', 20, data.vocab, d21.try_gpu())
     'it has it and the time tra'
gruModel128.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always the time traveller '
```

gruModel128.predict('he looked across', 20, data.vocab, d21.try gpu())

```
'he looked acrossed the provention is'
macs, params = get_model_complexity_info(gru128, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module GRU is treated as a zero-op.
      60.67 k, 100.000% Params, 63.05 MMac, 100.000% MACs,
      (rnn): GRU(60.67 k, 100.000% Params, 63.05 MMac, 100.000% MACs, 28, 128)
    Computational complexity: 63.05 MMac
    Number of parameters: 60.67 k
# Decreasing the number of hidden states to 8
gru8 = GRU(num_inputs=len(data.vocab), num_hiddens=8)
gruModel8 = d21.RNNLM(gru8, vocab_size=len(data.vocab), lr=4)
trainer.fit(gruModel8, data)
# Training time: 1min 29sec
                               train_ppl
      20
                              val_ppl
      15
      10
                    20
        0
              10
                          30
                                 40
                                       50
                      epoch
gruModel8.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has the the the the'
gruModel8.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always the the the the'
gruModel8.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked across the the the the'
macs, params = get_model_complexity_info(gru8, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module GRU is treated as a zero-op.
    GRU (
      912, 100.000% Params, 991.23 KMac, 100.000% MACs,
      (rnn): GRU(912, 100.000% Params, 991.23 KMac, 100.000% MACs, 28, 8)
    Computational complexity: 991.23 KMac
    Number of parameters: 912
1b) LSTM
class LSTM(d21.RNN):
    def __init__(self, num_inputs, num_hiddens):
        d21.Module.__init__(self)
        self.save_hyperparameters()
        self.rnn = nn.LSTM(num_inputs, num_hiddens)
```

```
def forward(self, inputs, H_C=None):
        return self.rnn(inputs, H C)
# Default from example
lstm = LSTM(num_inputs=len(data.vocab), num_hiddens=32)
lstmModel = d21.RNNLM(lstm, vocab_size=len(data.vocab), lr=4)
trainer.fit(lstmModel, data)
# Train time: 2min 30sec
                               train_ppl
      20
                             -- val_ppl
      15
      10
       5
        0
              10
                     20
                           30
                                 40
                                       50
                      epoch
lstmModel.predict('it has', 20, data.vocab, d21.try_gpu())
     'it has the traveller and t'
lstmModel.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always of the time travell'
lstmModel.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked acrossions and the time tr'
macs, params = get_model_complexity_info(lstm, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module LSTM is treated as a zero-op.
    LSTM(
      7.94 k, 100.000% Params, 8.45 MMac, 100.000% MACs,
       (rnn): LSTM(7.94 k, 100.000% Params, 8.45 MMac, 100.000% MACs, 28, 32)
     Computational complexity: 8.45 MMac
    Number of parameters: 7.94 k
# Halving the number of hidden states to 16
lstm16 = LSTM(num_inputs=len(data.vocab), num_hiddens=16)
LSTMmodel16 = d21.RNNLM(lstm16, vocab_size=len(data.vocab), lr=4)
trainer.fit(LSTMmodel16, data)
# Training time: 2min 3sec
                               train_ppl
      20
                               val ppl
     15
      10
```

20

epoch

30

40

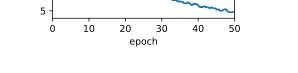
50

0

10

```
LSTMmodel16.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has the the the the'
LSTMmodel16.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always of the the the '
LSTMmodel16.predict('he looked across', 20, data.vocab, d21.try_gpu())
    'he looked across of the the the '
macs, params = get_model_complexity_info(lstm16, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module LSTM is treated as a zero-op.
    LSTM(
      2.94 k, 100.000% Params, 3.18 MMac, 100.000% MACs,
      (rnn): LSTM(2.94 k, 100.000% Params, 3.18 MMac, 100.000% MACs, 28, 16)
    Computational complexity: 3.18 MMac
    Number of parameters: 2.94 k
# Increasing the number of hidden states to 48
lstm48 = LSTM(num inputs=len(data.vocab), num hiddens=48)
LSTMmodel48 = d21.RNNLM(lstm48, vocab_size=len(data.vocab), lr=4)
trainer.fit(LSTMmodel48, data)
# Training time: 3min 1sec
                               train_ppl
      20
                              val_ppl
      15
      10
       5
        0
              10
                    20
                          30
                                 40
                                       50
                      epoch
LSTMmodel48.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has and the that the th'
LSTMmodel48.predict('we are always', 20, data.vocab, d21.try gpu())
    'we are always and the that the th'
LSTMmodel48.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked acrossed and the traveller'
macs, params = get_model_complexity_info(lstm48, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module LSTM is treated as a zero-op.
    LSTM(
      14.98 k, 100.000% Params, 15.83 MMac, 100.000% MACs,
       (rnn): LSTM(14.98 k, 100.000% Params, 15.83 MMac, 100.000% MACs, 28, 48)
    Computational complexity: 15.83 MMac
    Number of parameters: 14.98 k
```

```
# Increasing the number of hidden states to 64
lstm64 = LSTM(num_inputs=len(data.vocab), num_hiddens=64)
LSTMmodel64 = d21.RNNLM(lstm64, vocab size=len(data.vocab), lr=4)
trainer.fit(LSTMmodel64, data)
# Training time: 3min 54sec
                                train_ppl
      20
                               val_ppl
      15
      10
       5
              10
                     20
                           30
                                       50
        0
                                 40
                      epoch
LSTMmodel64.predict('it has', 20, data.vocab, d21.try_gpu())
     'it has inour and wather an'
LSTMmodel64.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always of and wathereard h'
LSTMmodel64.predict('he looked across', 20, data.vocab, d21.try_gpu())
     'he looked across have and the time t'
macs, params = get_model_complexity_info(lstm64, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
     Warning: module LSTM is treated as a zero-op.
      24.06 k, 100.000% Params, 25.3 MMac, 100.000% MACs,
      (rnn): LSTM(24.06 k, 100.000% Params, 25.3 MMac, 100.000% MACs, 28, 64)
     Computational complexity: 25.3 MMac
    Number of parameters: 24.06 k
# Increasing the number of hidden states to 128
lstm128 = LSTM(num_inputs=len(data.vocab), num_hiddens=128)
LSTMmodel128 = d21.RNNLM(lstm128, vocab size=len(data.vocab), lr=4)
trainer.fit(LSTMmodel128, data)
# Training time: 7min 32sec
                                train_ppl
      20
                                val_ppl
     15
      10
```



LSTMmodel128.predict('it has', 20, data.vocab, d21.try_gpu())

^{&#}x27;it has in and in and the t'

```
LSTMmodel128.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are alwayser and the the grome'
LSTMmodel128.predict('he looked across', 20, data.vocab, d2l.try_gpu())
    'he looked across of the thing the ti'
macs, params = get_model_complexity_info(lstm128, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module LSTM is treated as a zero-op.
    LSTM(
      80.9 k, 100.000% Params, 84.15 MMac, 100.000% MACs,
      (rnn): LSTM(80.9 k, 100.000% Params, 84.15 MMac, 100.000% MACs, 28, 128)
    Computational complexity: 84.15 MMac
    Number of parameters: 80.9 k
# Decreasing the number of hidden states to 8
lstm8 = LSTM(num_inputs=len(data.vocab), num_hiddens=8)
LSTMmodel8 = d21.RNNLM(lstm8, vocab_size=len(data.vocab), lr=4)
trainer.fit(LSTMmodel8, data)
# Training time: 2min
                               train_ppl
                               val_ppl
      20
      15
      10
        0
              10
                    20
                          30
                                 40
                                       50
                      epoch
LSTMmodel8.predict('it has', 20, data.vocab, d21.try gpu())
    'it has the the the the'
LSTMmodel8.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always and and and and'
LSTMmodel8.predict('he looked across', 20, data.vocab, d2l.try_gpu())
     'he looked across and and and and '
macs, params = get_model_complexity_info(lstm8, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
    Warning: module LSTM is treated as a zero-op.
      1.22 k, 100.000% Params, 1.33 MMac, 100.000% MACs,
      (rnn): LSTM(1.22 k, 100.000% Params, 1.33 MMac, 100.000% MACs, 28, 8)
    Computational complexity: 1.33 MMac
    Number of parameters: 1.22 k
```

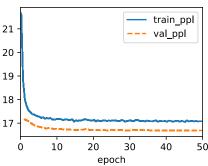
→ RNN Comparison

```
class RNN(d21.Module):
      """The RNN model implemented with high-level APIs."""
      def init (self, num inputs, num hiddens):
          super().__init__()
          self.save_hyperparameters()
          self.rnn = nn.RNN(num_inputs, num_hiddens)
      def forward(self, inputs, H=None):
          return self.rnn(inputs, H)
  rnn64 = RNN(num_inputs=len(data.vocab), num_hiddens=64)
  RNNmodel64 = d21.RNNLM(rnn64, vocab_size=len(data.vocab), lr=1)
  trainer.fit(RNNmodel64, data)
  # Training time: 2min 11sec
                                 train_ppl
        20
                                val_ppl
        15
        10
                       20
           0
                10
                             30
                                   40
                                         50
                        epoch
  RNNmodel64.predict('it has', 20, data.vocab, d21.try_gpu())
       'it has and hin the proun t'
  RNNmodel64.predict('we are always', 20, data.vocab, d21.try_gpu())
       'we are always all has in the time'
  RNNmodel64.predict('he looked across', 20, data.vocab, d21.try_gpu())
       'he looked across his said the time t'
  macs, params = get_model_complexity_info(rnn64, (1024, 28), as_strings=True,
                                             print_per_layer_stat=True, verbose=True)
  print('Computational complexity: ', macs)
  print('Number of parameters: ', params)
       Warning: module RNN is treated as a zero-op.
         6.02 k, 100.000% Params, 6.23 MMac, 100.000% MACs,
         (rnn): RNN(6.02 k, 100.000% Params, 6.23 MMac, 100.000% MACs, 28, 64)
       Computational complexity: 6.23 MMac
       Number of parameters: 6.02 k
2a) Deep GRU
  class deepGRU(d21.RNN):
      """The multi-layer GRU model."""
      def __init__(self, num_inputs, num_hiddens, num_layers, dropout=0.5):
          d21.Module.__init__(self)
          self.save_hyperparameters()
          self.rnn = nn.GRU(num_inputs, num_hiddens, num_layers,
                            dropout=dropout)
  # GRU w/ 3 layers
```

deepGru3 = deepGRU(num_inputs=len(data.vocab), num_hiddens=32, num_layers=3)
deepGRUModel3 = d21.RNNLM(deepGru3, vocab_size=len(data.vocab), lr=2)

```
trainer.fit(deepGRUModel3, data)
# Train time: 3min 14sec
```

```
macs, params = get_model_complexity_info(deepGru3, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
                                     _params__ are already defined for the moduleGRU ptflops can affect your code!
    Warning: variables __flops__ or _
    Warning: module deepGRU is treated as a zero-op.
    deepGRU(
      18.62 k, 100.000% Params, 19.76 MMac, 100.000% MACs,
      (rnn): GRU(18.62 k, 100.000% Params, 19.76 MMac, 100.000% MACs, 28, 32, num_layers=3, dropout=0.5)
    Computational complexity: 19.76 MMac
    Number of parameters: 18.62 k
deepGRUModel3.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has the the the the'
deepGRUModel3.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always the the the the'
deepGRUModel3.predict('he looked across', 20, data.vocab, d21.try gpu())
    'he looked across the the the the '
# GRU w/ 5 layers
deepGru5 = deepGRU(num inputs=len(data.vocab), num hiddens=32, num layers=5)
deepGRUModel5 = d21.RNNLM(deepGru5, vocab_size=len(data.vocab), lr=2)
trainer.fit(deepGRUModel5, data)
# Train time: 5min 30sec
                               train_ppl
```



Warning: module deepGRU is treated as a zero-op.

```
deepGRU(
    31.3 k, 100.000% Params, 33.19 MMac, 100.000% MACs,
    (rnn): GRU(31.3 k, 100.000% Params, 33.19 MMac, 100.000% MACs, 28, 32, num_layers=5, dropout=0.5)
)
Computational complexity: 33.19 MMac
Number of parameters: 31.3 k

deepGRUModel5.predict('it has', 20, data.vocab, d21.try_gpu())
    'it has

deepGRUModel5.predict('we are always', 20, data.vocab, d21.try_gpu())
    'we are always

deepGRUModel5.predict('he looked across', 20, data.vocab, d21.try_gpu())
    'he looked across
'
```

- 2b) Deep LSTM

```
class deepLSTM(d21.RNN):
    """The multi-layer LSTM model."""
    def __init__(self, num_inputs, num_hiddens, num_layers, dropout=0.5):
        d21.Module.__init__(self)
        self.save_hyperparameters()
        self.rnn = nn.LSTM(num_inputs, num_hiddens, num_layers,
                          dropout=dropout)
# LSTM w/ 3 layers
deepLSTM3 = deepLSTM(num_inputs=len(data.vocab), num_hiddens=32, num_layers=3)
deepLSTMModel3 = d21.RNNLM(deepLSTM3, vocab_size=len(data.vocab), lr=2)
trainer.fit(deepLSTMModel3, data)
# Train time: 4min 33sec
                               train_ppl
                              val_ppl
      22
      20
      18
              10
                     20
                           30
                                 40
                                       50
                      epoch
```

```
'we are always
deepLSTMModel3.predict('he looked across', 20, data.vocab, d21.try gpu())
     'he looked across
# LSTM w/ 5 layers
deepLSTM5 = deepLSTM(num inputs=len(data.vocab), num hiddens=32, num layers=5)
deepLSTMModel5 = d21.RNNLM(deepLSTM5, vocab_size=len(data.vocab), lr=2)
trainer.fit(deepLSTMModel5, data)
# Train time: 6min 37sec
      24 -
                               train_ppl
                              val_ppl
      22
      20
      18
                    20
                                       50
        0
              10
                           30
                                 40
                      epoch
macs, params = get_model_complexity_info(deepLSTM5, (1024, 28), as_strings=True,
                                           print_per_layer_stat=True, verbose=True)
print('Computational complexity: ', macs)
print('Number of parameters: ', params)
     Warning: module deepLSTM is treated as a zero-op.
     deepLSTM(
      41.73 k, 100.000% Params, 44.37 MMac, 100.000% MACs,
      (rnn): LSTM(41.73 k, 100.000% Params, 44.37 MMac, 100.000% MACs, 28, 32, num_layers=5, dropout=0.5)
     Computational complexity: 44.37 MMac
    Number of parameters: 41.73 k
deepLSTMModel5.predict('it has', 20, data.vocab, d21.try_gpu())
     'it has
deepLSTMModel5.predict('we are always', 20, data.vocab, d21.try_gpu())
     'we are always
deepLSTMModel5.predict('he looked across', 20, data.vocab, d21.try_gpu())
```

'he looked across

✓ 0s completed at 6:54 PM

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