```
!pip install setuptools==66
!pip install d2l==1.0.0-beta0
import cv2
import numpy as np
import pandas as pd
from tqdm import tqdm
from datetime import datetime
from matplotlib import pyplot as plt
import torch
from torch import nn
from d2l import torch as d2l
from torch.nn import functional as F
!pip install ptflops
import ptflops
from ptflops import get_model_complexity_info
import collections
import math
         Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>/
         Collecting setuptools==66
            Downloading setuptools-66.0.0-py3-none-any.whl (1.3 MB)
                                                                                       - 1.3/1.3 MB 14.1 MB/s eta 0:00:00
         Installing collected packages: setuptools
            Attempting uninstall: setuptools
                Found existing installation: setuptools 67.6.1
                Uninstalling setuptools-67.6.1:
                   Successfully uninstalled setuptools-67.6.1
         ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the
         ipython 7.34.0 requires jedi>=0.16, which is not installed.
         Successfully installed setuptools-66.0.0
         Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
        Collecting d2l==1.0.0-beta0
            Downloading d21-1.0.0b0-py3-none-any.whl (141 kB)
                                                                                      - 141.6/141.6 kB 5.0 MB/s eta 0:00:00
         Collecting gym==0.21.0
            Downloading gym-0.21.0.tar.gz (1.5 MB)
                                                                                       - 1.5/1.5 MB 39.1 MB/s eta 0:00:00
            Preparing metadata (setup.py) ... done
         Collecting gpytorch
            Downloading gpytorch-1.10-py3-none-any.whl (255 kB)
                                                                                    - 255.2/255.2 kB 27.0 MB/s eta 0:00:00
         Requirement already satisfied: pandas in /usr/local/lib/python3.9/dist-packages (from d2l==1.0.0-beta0) (1.5.3)
        Requirement already satisfied: requests in /usr/local/lib/python3.9/dist-packages (from d2l==1.0.0-beta0) (2.27.1)
         Requirement already satisfied: matplotlib-inline in /usr/local/lib/python3.9/dist-packages (from d2l==1.0.0-beta0) (0.1.6)
         Requirement already satisfied: scipy in /usr/local/lib/python 3.9/dist-packages (from d2l == 1.0.0-beta0) (1.10.1)
         Requirement already satisfied: matplotlib in /usr/local/lib/python3.9/dist-packages (from d2l==1.0.0-beta0) (3.7.1)
         Requirement already satisfied: numpy in /usr/local/lib/python3.9/dist-packages (from d2l==1.0.0-beta0) (1.22.4)
            Downloading jupyter-1.0.0-py2.py3-none-any.whl (2.7 kB)
         Requirement already satisfied: cloudpickle>=1.2.0 in /usr/local/lib/python3.9/dist-packages (from gym==0.21.0->d2l==1.0.0-beta0)
        Collecting linear-operator>=0.4.0
            Downloading linear_operator-0.4.0-py3-none-any.whl (156 kB)
                                                                                     - 156.7/156.7 kB 17.6 MB/s eta 0:00:00
         Requirement already satisfied: scikit-learn in /usr/local/lib/python 3.9/dist-packages (from gpytorch->d2l==1.0.0-beta0) (1.2.2)
         Requirement already satisfied: notebook in /usr/local/lib/python3.9/dist-packages (from jupyter->d2l==1.0.0-beta0) (6.4.8)
         Requirement already satisfied: ipykernel in /usr/local/lib/python3.9/dist-packages (from jupyter->d2l==1.0.0-beta0) (5.5.6)
         Requirement already satisfied: ipywidgets in /usr/local/lib/python3.9/dist-packages (from jupyter->d2l==1.0.0-beta0) (7.7.1)
         Requirement already satisfied: jupyter-console in /usr/local/lib/python3.9/dist-packages (from jupyter->d2l==1.0.0-beta0) (6.1.0
         Requirement already satisfied: nbconvert in /usr/local/lib/python3.9/dist-packages (from jupyter->d2l==1.0.0-beta0) (6.5.4)
         Collecting atconsole
            Downloading qtconsole-5.4.2-py3-none-any.whl (121 kB)
                                                                                     - 121.2/121.2 kB 14.9 MB/s eta 0:00:00
         Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (23
         Requirement \ already \ satisfied: \ python-date util>=2.7 \ in \ /usr/local/lib/python 3.9/dist-packages \ (from \ matplotlib->d2l==1.0.0-beta0) \ (from \ matplotl
         Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (0.11.6
         Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (1
         Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python 3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (in finite continuous formula for the finite continuous for the finite continuous formula for the finite continuous formula for the finite continuous formula for the finite continuous for the fi
         Requirement already satisfied: pillow>=6.2.0 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (8.4.0)
         Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (3
         Requirement already satisfied: kiwisolver>=1.0.1 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0-beta0) (
         Requirement already satisfied: importlib-resources>=3.2.0 in /usr/local/lib/python3.9/dist-packages (from matplotlib->d2l==1.0.0
         Requirement already satisfied: traitlets in /usr/local/lib/python3.9/dist-packages (from matplotlib-inline->d2l==1.0.0-beta0) (5
         Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.9/dist-packages (from pandas->d2l==1.0.0-beta0) (2022.7.1)
        Requirement already satisfied: urllib3<1.27,>=1.21.1 in /usr/local/lib/python3.9/dist-packages (from requests->d2l==1.0.0-beta0)
         Requirement already satisfied: charset-normalizer~=2.0.0 in /usr/local/lib/python3.9/dist-packages (from requests->d2l==1.0.0-be ▼
def init_seq2seq(module):
```

"""Initialize weights for Seq2Seq."""
if type(module) == nn.Linear:

nn.init.xavier\_uniform\_(module.weight)

```
for param in module._flat_weights_names:
            if "weight" in param:
                nn.init.xavier uniform (module. parameters[param])
class Seq2SeqEncoder(d21.Encoder):
      "The RNN encoder for sequence to sequence learning."""
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = d2l.GRU(embed_size, num_hiddens, num_layers, dropout)
        self.apply(init_seq2seq)
    def forward(self, X, *args):
        # X shape: (batch_size, num_steps)
        embs = self.embedding(X.t().type(torch.int64))
        # embs shape: (num_steps, batch_size, embed_size)
        outputs, state = self.rnn(embs)
        # outputs shape: (num_steps, batch_size, num_hiddens)
        # state shape: (num_layers, batch_size, num_hiddens)
        return outputs, state
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch_size, num_steps = 4, 9
encoder = d21.Seq2SeqEncoder(vocab_size, embed_size, num_hiddens, num_layers)
X = torch.zeros((batch_size, num_steps))
enc_outputs, enc_state = encoder(X)
d21.check_shape(enc_outputs, (num_steps, batch_size, num_hiddens))
d21.check_shape(enc_state, (num_layers, batch_size, num_hiddens))
class Seq2SeqDecoder(d21.Decoder):
      "The RNN decoder for sequence to sequence learning."""
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = d21.GRU(embed_size+num_hiddens, num_hiddens,
                          num_layers, dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(init_seq2seq)
    def init state(self, enc all outputs, *args):
        return enc_all_outputs
    def forward(self, X, state):
        # X shape: (batch_size, num_steps)
        # embs shape: (num_steps, batch_size, embed_size)
        embs = self.embedding(X.t().type(torch.int32))
        enc_output, hidden_state = state
        # context shape: (batch_size, num_hiddens)
        context = enc_output[-1]
        # Broadcast context to (num_steps, batch_size, num_hiddens)
        context = context.repeat(embs.shape[0], 1, 1)
        # Concat at the feature dimension
        embs and context = torch.cat((embs, context), -1)
        outputs, hidden_state = self.rnn(embs_and_context, hidden_state)
        outputs = self.dense(outputs).swapaxes(0, 1)
        # outputs shape: (batch_size, num_steps, vocab_size)
        # hidden_state shape: (num_layers, batch_size, num_hiddens)
        return outputs, [enc_output, hidden_state]
decoder = Seq2SeqDecoder(vocab_size, embed_size, num_hiddens, num_layers)
state = decoder.init_state(encoder(X))
dec_outputs, state = decoder(X, state)
d2l.check_shape(dec_outputs, (batch_size, num_steps, vocab_size))
d21.check_shape(state[1], (num_layers, batch_size, num_hiddens))
     /usr/local/lib/python3.9/dist-packages/torch/nn/modules/lazy.py:180: UserWarning: Lazy modules are a new feature under heavy develo
       warnings.warn('Lazy modules are a new feature under heavy development
class Seq2Seq(d21.EncoderDecoder):
```

"The RNN encoder-decoder for sequence to sequence learning."""

if type(module) == nn.GRU:

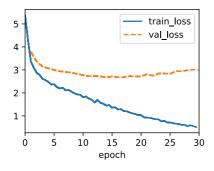
```
def __init__(self, encoder, decoder, tgt_pad, lr):
    super().__init__(encoder, decoder)
    self.save_hyperparameters()

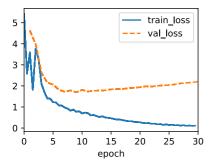
def validation_step(self, batch):
    Y_hat = self(*batch[:-1])
    self.plot('loss', self.loss(Y_hat, batch[-1]), train=False)

def configure_optimizers(self):
    # Adam optimizer is used here
    return torch.optim.Adam(self.parameters(), lr=self.lr)

@d21.add_to_class(Seq2Seq)
def loss(self, Y_hat, Y):
    l = super(Seq2Seq, self).loss(Y_hat, Y, averaged=False)
    mask = (Y.reshape(-1) != self.tgt_pad).type(torch.float32)
    return (1 * mask).sum() / mask.sum()
```

Adjust the hyperparameters to improve the translation results, embed\_size, num\_hiddens, num\_layers, dropout = 512, 512, 2, 0.2 lr= 0.oo1, Batch size= 256

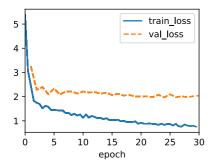




# question 1.2

```
self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = d21.GRU(embed_size, num_hiddens, num_layers, dropout)
        self.apply(init_seq2seq)
    def forward(self, X, *args):
        # X shape: (batch_size, num_steps)
        embs = self.embedding(X.t().type(torch.int64))
        # embs shape: (num_steps, batch_size, embed_size)
        outputs, state = self.rnn(embs)
        # outputs shape: (num_steps, batch_size, num_hiddens)
        # state shape: (num_layers, batch_size, num_hiddens)
        return outputs, state
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch size, num steps = 4, 9
encoder = Seq2SeqEncoder(vocab_size, embed_size, num_hiddens, num_layers)
X = torch.zeros((batch_size, num_steps))
enc\_outputs, enc\_state = encoder(X)
d21.check_shape(enc_outputs, (num_steps, batch_size, num_hiddens))
class Seq2SeqDecoder(d21.Decoder):
     ""The RNN decoder for sequence to sequence learning."""
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = d21.GRU(embed_size+num_hiddens, num_hiddens,
                           num_layers, dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(init_seq2seq)
    def init_state(self, enc_outputs, *args):
        enc_output, hidden_state = enc_outputs
        hidden_state = hidden_state.mean(dim=0, keepdim=True)
        hidden_state = hidden_state.repeat(self.rnn.num_layers, 1, 1)
        return enc_output, hidden_state
    def forward(self, X, state):
        # X shape: (batch_size, num_steps)
        # embs shape: (num_steps, batch_size, embed_size)
        embs = self.embedding(X.t().type(torch.int32))
        enc_output, hidden_state = state
        # context shape: (batch_size, num_hiddens)
        context = enc_output[-1]
        # Broadcast context to (num steps, batch size, num hiddens)
        context = context.repeat(embs.shape[0], 1, 1)
        # Concat at the feature dimension
        embs and context = torch.cat((embs, context), -1)
        outputs, hidden_state = self.rnn(embs_and_context, hidden_state)
        outputs = self.dense(outputs).swapaxes(0, 1)
        # outputs shape: (batch_size, num_steps, vocab_size)
        # hidden_state shape: (num_layers, batch_size, num_hiddens)
        return outputs, [enc_output, hidden_state]
decoder = Seq2SeqDecoder(vocab_size, embed_size, num_hiddens, num_layers)
state = decoder.init_state(encoder(X))
dec outputs, state = decoder(X, state)
d21.check_shape(dec_outputs, (batch_size, num_steps, vocab_size))
d21.check_shape(state[1], (num_layers, batch_size, num_hiddens))
embed_size, num_hiddens, num_layers, dropout = 384, 512, 3, 0.02
encoder = Seq2SeqEncoder(
   len(data.src_vocab), embed_size, num_hiddens, 3, dropout)
decoder = Seq2SeqDecoder(
    len(data.tgt_vocab), embed_size, num_hiddens, 2, dropout)
model = Seq2Seq(encoder, decoder, tgt_pad=data.tgt_vocab[''],
                1r=0.005)
trainer = d21.Trainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
```

```
train loss
                                val_loss
      1 .
LSTM MODEL
class LSTM(d21.RNN):
    def __init__(self, num_inputs, num_hiddens,num_layers=1,dropout=0):
        d21.Module.__init__(self)
        self.save_hyperparameters()
        self.rnn = nn.LSTM(num_inputs, num_hiddens,num_layers,dropout=dropout)
    def forward(self, inputs, H_C=None):
        return self.rnn(inputs, H_C)
class S2SEncoder LSTM(d21.Encoder):
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
        super().__init__()
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = LSTM(embed_size, num_hiddens, num_layers, dropout)
        self.apply(init_seq2seq)
    def forward(self, X, *args):
        embs = self.embedding(X.t().type(torch.int64))
        outputs, state = self.rnn(embs)
        return outputs, state
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch_size, num_steps = 4, 9
encoder = S2SEncoder_LSTM(vocab_size, embed_size, num_hiddens, num_layers)
X = torch.zeros((batch_size, num_steps))
enc_outputs, enc_state = encoder(X)
class S2SDecoder_LSTM(d21.Decoder):
    """The RNN decoder for sequence to sequence learning."""
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = LSTM(embed_size+num_hiddens, num_hiddens,
                          num_layers, dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(init_seq2seq)
    def init_state(self, enc_all_outputs, *args):
        return enc_all_outputs
    def forward(self, X, state):
        embs = self.embedding(X.t().type(torch.int32))
        enc_output, hidden_state = state
        context = enc_output[-1]
        context = context.repeat(embs.shape[0], 1, 1)
        embs_and_context = torch.cat((embs, context), -1)
        outputs, hidden_state = self.rnn(embs_and_context, hidden_state)
        outputs = self.dense(outputs).swapaxes(0, 1)
        return outputs, [enc_output, hidden_state]
decoder = S2SDecoder_LSTM(vocab_size, embed_size, num_hiddens, num_layers)
state = decoder.init_state(encoder(X))
dec_outputs, state = decoder(X, state)
d21.check_shape(dec_outputs, (batch_size, num_steps, vocab_size))
embed_size, num_hiddens, num_layers, dropout = 384, 512, 3, 0.3
encoder = S2SEncoder_LSTM(
   len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = S2SDecoder_LSTM(
```



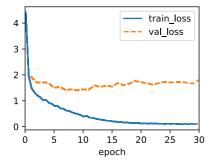
### Q:2- Explore the impacts of number of hidden layers starting tom 1 hidden layer up to 4 hidden layers.

#### In Class Example with 1 hidden layers

```
class AttentionDecoder(d21.Decoder):
     ""The base attention-based decoder interface."""
    def __init__(self):
        super().__init__()
    @property
    def attention_weights(self):
        raise NotImplementedError
class Seq2SeqAttentionDecoder(AttentionDecoder):
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                 dropout=0):
        super().__init__()
        self.attention = d21.AdditiveAttention(num_hiddens, dropout)
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = nn.GRU(
            embed_size + num_hiddens, num_hiddens, num_layers,
            dropout=dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(d21.init_seq2seq)
    def init_state(self, enc_outputs, enc_valid_lens):
        outputs, hidden_state = enc_outputs
        return (outputs.permute(1, 0, 2), hidden_state, enc_valid_lens)
    def forward(self, X, state):
        enc_outputs, hidden_state, enc_valid_lens = state
        X = self.embedding(X).permute(1, 0, 2)
        outputs, self._attention_weights = [], []
        for x in X:
            query = torch.unsqueeze(hidden_state[-1], dim=1)
            context = self.attention(
                query, enc_outputs, enc_outputs, enc_valid_lens)
            x = torch.cat((context, torch.unsqueeze(x, dim=1)), dim=-1)
            out, hidden_state = self.rnn(x.permute(1, 0, 2), hidden_state)
            outputs.append(out)
            self._attention_weights.append(self.attention.attention_weights)
        outputs = self.dense(torch.cat(outputs, dim=0))
        return\ outputs.permute (1,\ 0,\ 2),\ [enc\_outputs,\ hidden\_state,
                                          enc_valid_lens]
    @property
    def attention_weights(self):
        return self._attention_weights
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch_size, num_steps = 4, 7
encoder = d21.Seq2SeqEncoder(vocab_size, embed_size, num_hiddens, num_layers)
decoder = Seq2SeqAttentionDecoder(vocab_size, embed_size, num_hiddens,
                                  num_layers)
X = torch.zeros((batch_size, num_steps), dtype=torch.long)
state = decoder.init_state(encoder(X), None)
output, state = decoder(X, state)
dol chack chang(nutnut (hatch size num stens youch size))
```

```
uzi.cncck_snapc(oucpuc, (ouccn_size, nam_sceps, vocab_size))
d21.check_shape(state[0], (batch_size, num_steps, num_hiddens))
d21.check_shape(state[1][0], (batch_size, num_hiddens))
data = d21.MTFraEng(batch_size=128)
embed_size, num_hiddens, num_layers, dropout = 256, 256, 1, 0.2
encoder = d21.Seq2SeqEncoder(
       len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = Seq2SeqAttentionDecoder(
        len(data.tgt vocab), embed size, num hiddens, num layers, dropout)
model = d21.Seq2Seq(encoder, decoder, tgt_pad=data.tgt_vocab[''],
                                        lr=0.005)
trainer = d21.Trainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
                                                               train_loss
                                                         --- val loss
             3
             2
             1
             0
                                      10
                                                 15
                                                             20
                                                                        25
                                              epoch
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
        data.build(engs, fras), d21.try_gpu(), data.num_steps)
for en, fr, p in zip(engs, fras, preds):
        translation = []
        for token in data.tgt_vocab.to_tokens(p):
               if token == '':
                       break
                translation.append(token)
        print(f'{en} => {translation}, bleu,'
                    f'\{d21.bleu("\ ".join(translation),\ fr,\ k=2):.3f\}')
_, dec_attention_weights = model.predict_step(
        data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
        [step[0][0][0] \ for \ step \ in \ dec_attention\_weights], \ 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
        attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
        xlabel='Key positions', ylabel='Query positions')
         go . => ['va', '!', '<eos>', '<pad>', '<pad
                 0
            bositions 4 -
                                                  0.75
                                                  0.50
             Query
                                                  0.25
                 6 -
                  8
                        0
                     Key positions
With 2 hidden layers
data = d21.MTFraEng(batch_size=128)
embed_size, num_hiddens, num_layers, dropout = 256, 256, 2, 0.2
encoder = d21.Seq2SeqEncoder(
       len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = Seq2SeqAttentionDecoder(
       len(data.tgt_vocab), embed_size, num_hiddens, num_layers, dropout)
model = d21.Seq2Seq(encoder, decoder, tgt_pad=data.tgt_vocab[''],
                    lr=0.005)
```

```
trainer = d21.Irainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
```



```
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
            data.build(engs, fras), d21.try_gpu(), data.num_steps)
 for en, fr, p in zip(engs, fras, preds):
             translation = []
             for token in data.tgt_vocab.to_tokens(p):
                         if token == '':
                                      break
                         translation.append(token)
             print(f'{en} => {translation}, bleu,'
                              f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
         dec_attention_weights = model.predict_step(
             data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
             [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
             attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
             xlabel='Key positions', ylabel='Query positions')
               go . => ['va', '!', '<eos>', '<pad>', '<pad
                            0
                   y positions
                                                                               0.75
                                                                                0.50
                    Query F
                                                                                0.25
                                     0
                                  Key positions
```

## With 3 hidden layers

```
- train loss
                    4 -
                                                                                        --- val_loss
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
            data.build(engs, fras), d21.try_gpu(), data.num_steps)
for en, fr, p in zip(engs, fras, preds):
             translation = []
            for token in data.tgt_vocab.to_tokens(p):
                         if token == '':
                                    break
                         {\tt translation.append(token)}
            print(f'{en} => {translation}, bleu,'
                                f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
_, dec_attention_weights = model.predict_step(
            data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
            [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
             attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
            xlabel='Key positions', ylabel='Query positions')
               go . => ['va', '!', '<eos>', '<pad>', '<pad
                           0
                   Query positions
9 6 0
                                                                              0.75
                                                                               0.50
                                                                              0.25
                            8 -
                                     0
                                                      2
                                 Key positions
```

### With 4 hidden layers

```
train_loss

val_loss

3 -

2 -

1 -

0 5 10 15 20 25 30

epoch
```

```
translation.append(token)
                print(f'{en} => {translation}, bleu,'
                                          f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
   , dec attention weights = model.predict step(
                data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
                [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
                 attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
                xlabel='Key positions', ylabel='Query positions')
                    go . => ['va', '!', '<eos>', '<pad>', '<pad
                           bositions 4 -
                                                                                                    - 0.75
                                                                                                      0.50
                           Query 1
                                    8 -
                                            Key positions
```

Replace GRU with LSTM in the experiment. Perform training again. Plot the results

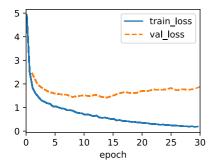
#### With 1 Hidden layer

```
class Seq2SeqAttentionDecoder_LSTM(AttentionDecoder):
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.attention = d21.AdditiveAttention(num_hiddens, dropout)
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = nn.LSTM(
            embed_size + num_hiddens, num_hiddens, num_layers,
            dropout=dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(d21.init_seq2seq)
    def init_state(self, enc_outputs, enc_valid_lens):
        outputs, hidden state = enc outputs
        cell_state = hidden_state.new_zeros(hidden_state.shape)
        return (outputs.permute(1, 0, 2), (hidden_state, cell_state), enc_valid_lens)
    def forward(self, X, state):
        enc_outputs, hidden_and_cell_state, enc_valid_lens = state
        X = self.embedding(X).permute(1, 0, 2)
        outputs, self._attention_weights = [], []
        for x in X:
            query = torch.unsqueeze(hidden_and_cell_state[0][-1], dim=1)
            context = self.attention(
               query, enc_outputs, enc_outputs, enc_valid_lens)
            x = torch.cat((context, torch.unsqueeze(x, dim=1)), dim=-1)
            out, hidden_and_cell_state = self.rnn(x.permute(1, 0, 2), hidden_and_cell_state)
            outputs.append(out)
            self._attention_weights.append(self.attention.attention_weights)
        outputs = self.dense(torch.cat(outputs, dim=0))
        return outputs.permute(1, 0, 2), [enc_outputs, hidden_and_cell_state,
                                          enc_valid_lens]
    @property
    def attention_weights(self):
        return self. attention weights
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch_size, num_steps = 4, 7
encoder = d21.Seq2SeqEncoder(vocab_size, embed_size, num_hiddens, num_layers)
decoder = Seq2SeqAttentionDecoder_LSTM(vocab_size, embed_size, num_hiddens,
                                  num_layers)
X = torch.zeros((batch_size, num_steps), dtype=torch.long)
state = decoder.init_state(encoder(X), None)
output, state = decoder(X, state)
```

```
d21.check_shape(output, (batch_size, num_steps, vocab_size))
d21.check_shape(state[0], (batch_size, num_steps, num_hiddens))
data = d21.MTFraEng(batch_size=128)
embed_size, num_hiddens, num_layers, dropout = 256, 256, 1, 0.2
encoder = d21.Seq2SeqEncoder(
         len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = Seq2SeqAttentionDecoder_LSTM(
         len(data.tgt_vocab), embed_size, num_hiddens, num_layers, dropout)
model = d21.Seq2Seq(encoder, decoder, tgt_pad=data.tgt_vocab[''],
                                              1r=0.005)
trainer = d21.Trainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
                                                                          train_loss
               4
                                                                     -- val_loss
               3
               2
               1
                                            10
                                                        15
                                                                      20
                                                                                   25
                                                                                                30
                                                      epoch
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
         data.build(engs, fras), d21.try_gpu(), data.num_steps)
for en, fr, p in zip(engs, fras, preds):
         translation = []
         for token in data.tgt_vocab.to_tokens(p):
                  if token == '':
                           break
                  {\tt translation.append(token)}
         print(f'{en} => {translation}, bleu,'
                       f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
_, dec_attention_weights = model.predict_step(
         data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
         [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
         attention\_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),\\
         xlabel='Key positions', ylabel='Query positions')
           go . => ['va', '!', '<eos>', '<pad>', '<pad
                    0
               Query positions 9 4 6
                                                          0.4
                                                          0.2
                                        2
                        Key positions
```

### With 2 Hidden layers

```
trainer = d21.Trainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
```

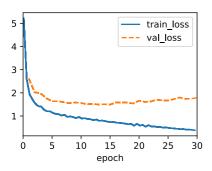


```
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
           data.build(engs, fras), d21.try_gpu(), data.num_steps)
 for en, fr, p in zip(engs, fras, preds):
            translation = []
            for token in data.tgt_vocab.to_tokens(p):
                        if token == '':
                                  break
                        translation.append(token)
            print(f'{en} => {translation}, bleu,'
                              f'{d2l.bleu(" ".join(translation), fr, k=2):.3f}')
_, dec_attention_weights = model.predict_step(
           data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
            [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
            attention\_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),\\
            xlabel='Key positions', ylabel='Query positions')
              go . => ['va', '!', '<eos>', '<pad>', '<pad
                          0
                  y positions
A
                                                                            0.4
                                                                            0.2
                    Query
                         6
                                    0
                                Key positions
```

#### With 3 Hidden layers

```
train loss
                                                                                         --- val_loss
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
            data.build(engs, fras), d21.try_gpu(), data.num_steps)
 for en, fr, p in zip(engs, fras, preds):
            translation = []
            for token in data.tgt_vocab.to_tokens(p):
                         if token == '':
                                     break
                         translation.append(token)
            print(f'{en} => {translation}, bleu,'
                               f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
         dec_attention_weights = model.predict_step(
            data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
 attention_weights = torch.cat(
            [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
             attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
            xlabel='Key positions', ylabel='Query positions')
               go . => ['<unk>', '.', '<eos>', '<pad>', '<
                            0
                     v positions
                                                                               0.4
                                                                               0.3
                     Query
                           6
                                                                              0.2
                           8
                                     0
                                                      2
                                 Key positions
```

### With 4 Hidden layers



```
translation.append(token)
                print(f'{en} => {translation}, bleu,'
                                          f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
  _, dec_attention_weights = model.predict_step(
                data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
                [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
                 attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
                xlabel='Key positions', ylabel='Query positions')
                    go . => ['<unk>', 'à', 'la', 'maison', '.', '<eos>', '<pad>', '<pad>', '<pad>'], b
i lost . => ['je', "l'ai", 'vu', '.', '<eos>', '<pad>', '<pad>', '<pad>', '<pad>', '<pad>']
he's calm . => ['il', 'est', '<unk>', '.', '<eos>', '<pad>', '<pad
                                    0
                           bositions 4
                                                                                                       0.3
                           Query p
                                                                                                     0.2
                                    8
                                                                       2
                                            Key positions
```

#### **Bonus Question**

### With 1 Hidden layers

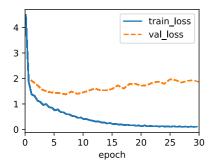
```
class Seq2SeqAttentionDecoder(AttentionDecoder):
    def __init__(self, vocab_size, embed_size, num_hiddens, num_layers,
                dropout=0):
        super().__init__()
        self.attention = d21.DotProductAttention(dropout)
        self.embedding = nn.Embedding(vocab_size, embed_size)
        self.rnn = nn.GRU(
            embed_size + num_hiddens, num_hiddens, num_layers,
            dropout=dropout)
        self.dense = nn.LazyLinear(vocab_size)
        self.apply(d21.init_seq2seq)
    def init_state(self, enc_outputs, enc_valid_lens):
        outputs, hidden state = enc outputs
        return (outputs.permute(1, 0, 2), hidden_state, enc_valid_lens)
    def forward(self, X, state):
        enc_outputs, hidden_state, enc_valid_lens = state
        X = self.embedding(X).permute(1, 0, 2)
        outputs, self._attention_weights = [], []
        for x in X:
            query = torch.unsqueeze(hidden_state[-1], dim=1)
            context = self.attention(
                query, enc_outputs, enc_outputs, enc_valid_lens)
            x = torch.cat((context, torch.unsqueeze(x, dim=1)), dim=-1)
           out, hidden_state = self.rnn(x.permute(1, 0, 2), hidden_state)
           outputs.append(out)
            self._attention_weights.append(self.attention.attention_weights)
        outputs = self.dense(torch.cat(outputs, dim=0))
        return outputs.permute(1, 0, 2), [enc_outputs, hidden_state,
                                          enc valid lens]
    @property
    def attention_weights(self):
        return self._attention_weights
vocab_size, embed_size, num_hiddens, num_layers = 10, 8, 16, 2
batch_size, num_steps = 4, 7
encoder = d21.Seq2SeqEncoder(vocab_size, embed_size, num_hiddens, num_layers)
decoder = Seq2SeqAttentionDecoder(vocab_size, embed_size, num_hiddens,
                                  num_layers)
X = torch.zeros((batch_size, num_steps), dtype=torch.long)
state = decoder.init_state(encoder(X), None)
```

```
output, state = decoder(X, state)
d21.check_shape(output, (batch_size, num_steps, vocab_size))
d21.check_shape(state[0], (batch_size, num_steps, num_hiddens))
d21.check_shape(state[1][0], (batch_size, num_hiddens))
data = d21.MTFraEng(batch size=128)
embed_size, num_hiddens, num_layers, dropout = 256, 256, 1, 0.2
encoder = d21.Seq2SeqEncoder(
    len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = Seq2SeqAttentionDecoder(
    len(data.tgt_vocab), embed_size, num_hiddens, num_layers, dropout)
model = d21.Seq2Seq(encoder, decoder, tgt_pad=data.tgt_vocab[''],
                     1r=0.005)
trainer = d21.Trainer(max_epochs=30, gradient_clip_val=1, num_gpus=1)
trainer.fit(model, data)
       4
                                   train_loss
                                 val_loss
       3
       2
       1
       0
                    10
                          15
                                20
                                      25
        0
                                            30
                        epoch
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
    data.build(engs, fras), d21.try_gpu(), data.num_steps)
for en, fr, p in zip(engs, fras, preds):
    translation = []
    for token in data.tgt_vocab.to_tokens(p):
        if token == '':
            break
        translation.append(token)
    print(f'{en} \Rightarrow {translation}, bleu,'
          f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
```

```
dec_attention_weights = model.predict_step(
                    data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
 attention_weights = torch.cat(
                    [step[0][0][0] \ for \ step \ in \ dec_attention\_weights], \ 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
 d21.show heatmaps(
                   attention\_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),\\
                   xlabel='Key positions', ylabel='Query positions')
                       go . => ['va', '!', '<eos>', '<pad>', '<pad
                                          0
                               positions
4
                                                                                                                         0.4
                                                                                                                         0.2
                               Query
9
                                           8
                                                         0
                                                                                   2
                                                   Key positions
```

## With 2 Hidden layers

```
data = d21.MTFraEng(batch_size=128)
embed_size, num_hiddens, num_layers, dropout = 256, 256, 2, 0.2
encoder = d21.Seq2SeqEncoder(
    len(data.src_vocab), embed_size, num_hiddens, num_layers, dropout)
decoder = Seq2SeqAttentionDecoder(
    len(data.tgt_vocab), embed_size, num_hiddens, num_layers, dropout)
```

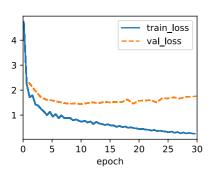


```
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
            data.build(engs, fras), d21.try_gpu(), data.num_steps)
 for en, fr, p in zip(engs, fras, preds):
             translation = []
             for token in data.tgt_vocab.to_tokens(p):
                         if token == '':
                                     break
                         translation.append(token)
             print(f'{en} => {translation}, bleu,'
                                f'{d2l.bleu(" ".join(translation), fr, k=2):.3f}')
         dec_attention_weights = model.predict_step(
             \verb|data.build([engs[-1]], [fras[-1]]), | d2l.try\_gpu(), | data.num\_steps, | True)|
 attention_weights = torch.cat(
             [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show heatmaps(
             attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
             xlabel='Key positions', ylabel='Query positions')
               go . => ['va', '!', '<eos>', '<pad>', '<pad
                            0
                    v positions
                                                                                 0.3
                                                                               0.2
                     Query
                                                                                0.1
                           6
                            8 -
                                      0
                                                       2
                                  Key positions
```

### With 3 Hidden layers

```
train loss
                     4
                                                                                                   - val_loss
                     3
                     2
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
            data.build(engs, fras), d21.try_gpu(), data.num_steps)
 for en, fr, p in zip(engs, fras, preds):
             translation = []
            for token in data.tgt_vocab.to_tokens(p):
    if token == '':
                                    break
                          {\tt translation.append(token)}
             print(f'{en} => {translation}, bleu,'
                                f'{d2l.bleu(" ".join(translation), fr, k=2):.3f}')
_, dec_attention_weights = model.predict_step(
            data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
             [step[0][0][0] for step in dec_attention_weights], 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
             attention\_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),\\
             xlabel='Key positions', ylabel='Query positions')
               go . => ['va', '!', '<eos>', '<pad>', '<pad
                            0
                     Query positions 6
                                                                                  0.4
                             8
                                  Key positions
```

## With 4 Hidden layers



```
engs = ['go .', 'i lost .', 'he\'s calm .', 'i\'m home .']
fras = ['va !', 'j\'ai perdu .', 'il est calme .', 'je suis chez moi .']
preds, _ = model.predict_step(
    data.build(engs, fras), d21.try_gpu(), data.num_steps)
```

```
for en, fr, p in zip(engs, fras, preds):
              translation = []
              for token in data.tgt_vocab.to_tokens(p):
                             if token == '':
                                           break
                             translation.append(token)
              print(f'\{en\} \Rightarrow \{translation\}, bleu,'
                                    f'{d21.bleu(" ".join(translation), fr, k=2):.3f}')
_, dec_attention_weights = model.predict_step(
              data.build([engs[-1]], [fras[-1]]), d2l.try_gpu(), data.num_steps, True)
attention_weights = torch.cat(
              [step[0][0][0] \ for \ step \ in \ dec_attention\_weights], \ 0)
attention_weights = attention_weights.reshape((1, 1, -1, data.num_steps))
d21.show_heatmaps(
              attention_weights[:, :, :, :len(engs[-1].split()) + 1].cpu(),
              xlabel='Key positions', ylabel='Query positions')
                 go . => ['va', '!', '<eos>', '<pad>', '<pad
                               0
                       Query positions
                                                                                          0.4
                                                                                           0.2
                               8
                                       Key positions
```