

# Capstone Project

## Neural translation model

### Instructions

In this notebook, you will create a neural network that translates from English to German. You will use concepts from throughout this course, including building more flexible model architectures, freezing layers, data processing pipeline and sequence modelling.

This project is peer-assessed. Within this notebook you will find instructions in each section for how to complete the project. Pay close attention to the instructions as the peer review will be carried out according to a grading rubric that checks key parts of the project instructions. Feel free to add extra cells into the notebook as required.

### How to submit

When you have completed the Capstone project notebook, you will submit a pdf of the notebook for peer review. First ensure that the notebook has been fully executed from beginning to end, and all of the cell outputs are visible. This is important, as the grading rubric depends on the reviewer being able to view the outputs of your notebook. Save the notebook as a pdf (you could download the notebook with File -> Download .ipynb, open the notebook locally, and then File -> Download as -> PDF via LaTeX), and then submit this pdf for review.

### Let's get started!

We'll start by running some imports, and loading the dataset. For this project you are free to make further imports throughout the notebook as you wish.

In [2]:

```
import tensorflow as tf
import tensorflow_hub as hub
import unicodedata
import re
from IPython.display import Image
```

For the capstone project, you will use a language dataset from <http://www.manythings.org/anki/> to build a neural translation model. This dataset consists of over 200,000 pairs of sentences in English and German. In order to make the training quicker, we will restrict to our dataset to 20,000 pairs. Feel free to change this if you wish - the size of the dataset used is not part of the grading rubric.

Your goal is to develop a neural translation model from English to German, making use of a pre-trained English word embedding module.

### Import the data

The dataset is available for download as a zip file at the following link:

<https://drive.google.com/open?id=1KczOciG7sYY7SB9UIBeRP1T9659b121Q>

You should store the unzipped folder in Drive for use in this Colab notebook.

In [3]:

```
# Run this cell to connect to your Drive folder

# from google.colab import drive
# drive.mount('/content/gdrive')

from google.colab import drive
drive.mount('/content/drive')

import os
path = '/content/drive/My Drive/INSAID/TensorFlow/Customizing your models with Tensorflow 2/Week5'
os.chdir(path)
```

```
os.listdir()
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force\_remount=True).

Out[3]:

```
['deu.txt',  
 'english-german.pkl',  
 'encoder.h5',  
 'decoder.h5',  
 'Capstone_Project.ipynb',  
 'nmt_with_attention.ipynb',  
 'machine_translation_tensorflow_orig.ipynb',  
 'neural_translation_model.png']
```

In [4]:

```
# Run this cell to load the dataset  
  
NUM_EXAMPLES = 20000  
data_examples = []  
with open(os.path.join(path, 'deu.txt'), 'r', encoding='utf8') as f:  
    for line in f.readlines():  
        if len(data_examples) < NUM_EXAMPLES:  
            data_examples.append(line)  
        else:  
            break
```

In [5]:

```
# These functions preprocess English and German sentences  
  
def unicode_to_ascii(s):  
    return ''.join(c for c in unicodedata.normalize('NFD', s) if unicodedata.category(c) != 'Mn')  
  
def preprocess_sentence(sentence):  
    sentence = sentence.lower().strip()  
    sentence = re.sub(r"ü", 'ue', sentence)  
    sentence = re.sub(r"ä", 'ae', sentence)  
    sentence = re.sub(r"ö", 'oe', sentence)  
    sentence = re.sub(r"ß", 'ss', sentence)  
  
    sentence = unicode_to_ascii(sentence)  
    sentence = re.sub(r"([?.,!])", r" \1 ", sentence)  
    sentence = re.sub(r"^[^a-z?.,!]+", " ", sentence)  
    sentence = re.sub(r'[" "]+' , " ", sentence)  
  
    return sentence.strip()
```

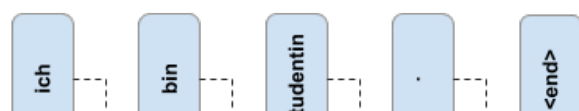
## The custom translation model

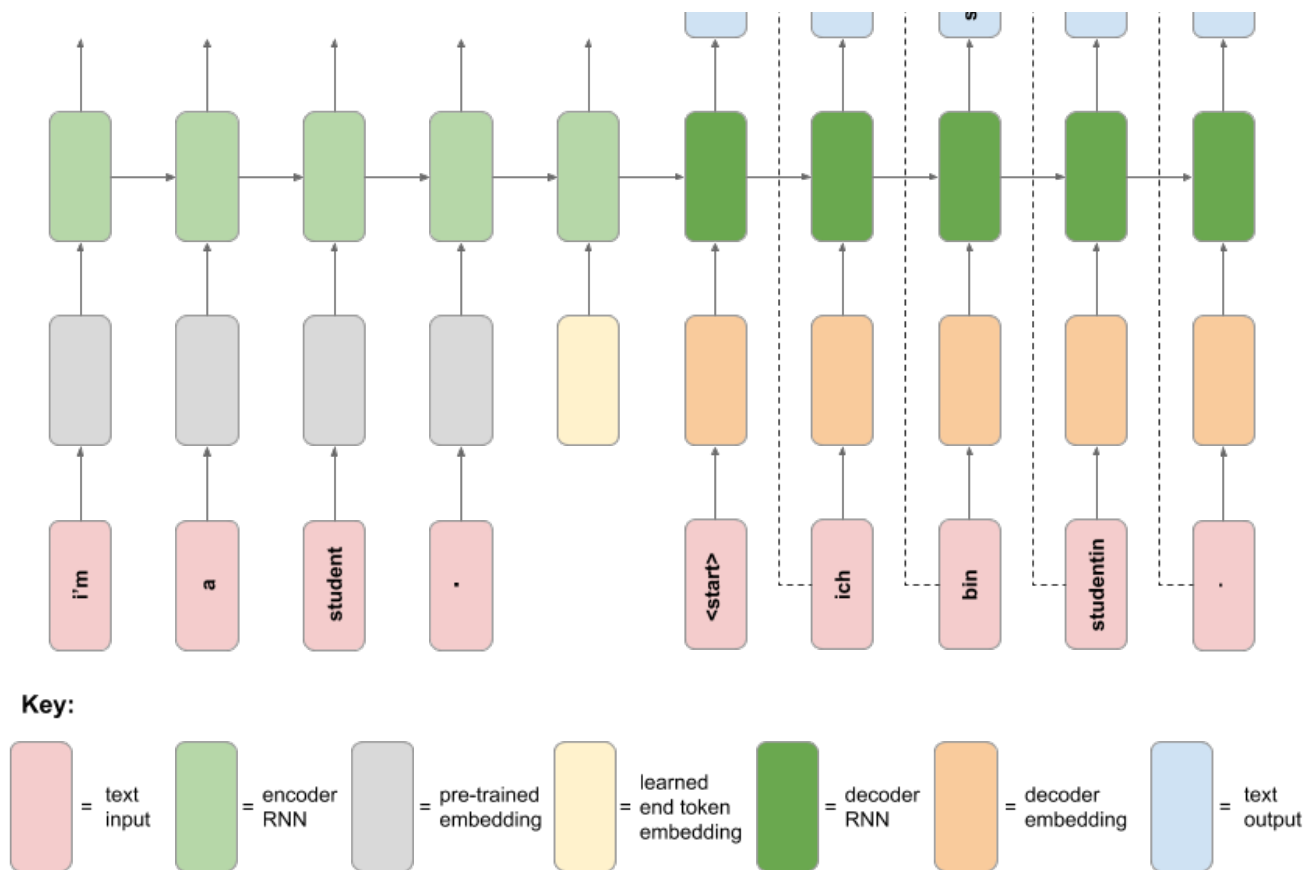
The following is a schematic of the custom translation model architecture you will develop in this project.

In [6]:

```
# Run this cell to download and view a schematic diagram for the neural translation model  
  
!wget -q -O neural_translation_model.png --no-check-certificate "https://docs.google.com/uc?export=download&id=1XsS1VlXoaEo-RbYNilJ9jcscNZvsSPmd"  
Image("neural_translation_model.png")
```

Out[6]:





The custom model consists of an encoder RNN and a decoder RNN. The encoder takes words of an English sentence as input, and uses a pre-trained word embedding to embed the words into a 128-dimensional space. To indicate the end of the input sentence, a special end token (in the same 128-dimensional space) is passed in as an input. This token is a TensorFlow Variable that is learned in the training phase (unlike the pre-trained word embedding, which is frozen).

The decoder RNN takes the internal state of the encoder network as its initial state. A start token is passed in as the first input, which is embedded using a learned German word embedding. The decoder RNN then makes a prediction for the next German word, which during inference is then passed in as the following input, and this process is repeated until the special `<end>` token is emitted from the decoder.

## 1. Text preprocessing

- Create separate lists of English and German sentences, and preprocess them using the `preprocess_sentence` function provided for you above.
- Add a special `"<start>"` and `"<end>"` token to the beginning and end of every German sentence.
- Use the `Tokenizer` class from the `tf.keras.preprocessing.text` module to tokenize the German sentences, ensuring that no character filters are applied. *Hint: use the `Tokenizer`'s `"filter"` keyword argument.*
- Print out at least 5 randomly chosen examples of (preprocessed) English and German sentence pairs. For the German sentence, print out the text (with start and end tokens) as well as the tokenized sequence.
- Pad the end of the tokenized German sequences with zeros, and batch the complete set of sequences into one numpy array.

In [7]:

```
# Inspect the first few data examples
data_examples[:5]
```

Out[7]:

```
['Hi.\tHallo!\tCC-BY 2.0 (France) Attribution: tatoeba.org #538123 (CM) & #380701 (cburgmer)\n',
 'Hi.\tGrüß Gott!\tCC-BY 2.0 (France) Attribution: tatoeba.org #538123 (CM) & #659813 (Esperantostern)\n',
 'Run!\tLauf!\tCC-BY 2.0 (France) Attribution: tatoeba.org #906328 (papabear) & #941078 (Fingerhut)\n',
 'Wow!\tPötzdonner!\tCC-BY 2.0 (France) Attribution: tatoeba.org #52027 (Zifre) & #2122382 (Pfirsichbaeumchen)\n',
 'Wow!\tDonnerwetter!\tCC-BY 2.0 (France) Attribution: tatoeba.org #52027 (Zifre) & #2122391 (Pfirsichbaeumchen)\n']
```

In [8]:

```
# Extract the English and the German sentences
english = [l.split("\t")[0] for l in data_examples]
german = [l.split("\t")[1] for l in data_examples]
```

In [9]:

```
# Preprocess the language datasets
import numpy as np

english = np.array([preprocess_sentence(s) for s in english])
german = np.array([preprocess_sentence(s) for s in german])
```

In [10]:

```
# Add start and end tokens
german = np.array(["<start> {} <end>".format(s) for s in german])
```

In [11]:

```
# Tokenize the German sequences
from tensorflow.keras.preprocessing.text import Tokenizer

german_tokenizer = Tokenizer(filters='')
german_tokenizer.fit_on_texts(german)
```

In [12]:

```
# Convert the German sentences to token sequences
tokenized_german_sequences = german_tokenizer.texts_to_sequences(german)
```

In [13]:

```
# Inspect a few examples of each
inx = np.random.choice(len(english), 5, replace=False)

for eng, ger in zip(english[inx], german[inx]):
    print(eng)
    print(ger)
    print(german_tokenizer.texts_to_sequences(ger.split()))
    print('')
```

i am out of work .

<start> ich bin arbeitslos . <end>

[[1], [4], [15], [2051], [3], [2]]

make it stop .

<start> mach , dass es aufhoert ! <end>

[[1], [79], [25], [132], [10], [2482], [9], [2]]

we miss you , too .

<start> wir vermissen dich auch . <end>

[[1], [17], [473], [28], [112], [3], [2]]

she helps us .

<start> sie hilft uns . <end>

[[1], [8], [545], [31], [3], [2]]

use your fist .

<start> nutze deine faust ! <end>

[[1], [1930], [117], [2557], [9], [2]]

In [14]:

```
# Pad the tokenized sequences with zeros
from tensorflow.keras.preprocessing.sequence import pad_sequences

padded_german_seq = pad_sequences(tokenized_german_sequences, padding="post")
```

In [15]:

```
# Get the number of tokens
num_german_tokens = max(german_tokenizer.index_word.keys()) + 1
print(num_german_tokens)
```

5744

## 2. Prepare the data

### Load the embedding layer

As part of the dataset preprocessing for this project, you will use a pre-trained English word embedding module from TensorFlow Hub. The URL for the module is <https://tfhub.dev/google/tf2-preview/nnlm-en-dim128-with-normalization/1>.

This embedding takes a batch of text tokens in a 1-D tensor of strings as input. It then embeds the separate tokens into a 128-dimensional space.

The code to load and test the embedding layer is provided for you below.

**NB:** this model can also be used as a sentence embedding module. The module will process each token by removing punctuation and splitting on spaces. It then averages the word embeddings over a sentence to give a single embedding vector. However, we will use it only as a word embedding module, and will pass each word in the input sentence as a separate token.

In [16]:

```
# Load embedding module from Tensorflow Hub
embedding_layer = hub.KerasLayer("https://tfhub.dev/google/tf2-preview/nnlm-en-dim128/1",
                                output_shape=[128], input_shape=[], dtype=tf.string)
```

In [17]:

```
# Test the layer
embedding_layer(tf.constant(["these", "aren't", "the", "droids", "you're", "looking", "for"])).shape
```

Out[17]:

TensorShape([7, 128])

You should now prepare the training and validation Datasets.

- Create a random training and validation set split of the data, reserving e.g. 20% of the data for validation (NB: each English dataset example is a single sentence string, and each German dataset example is a sequence of padded integer tokens).
- Load the training and validation sets into a `tf.data.Dataset` object, passing in a tuple of English and German data for both training and validation sets.
- Create a function to map over the datasets that splits each English sentence at spaces. Apply this function to both Dataset objects using the map method. *Hint: look at the `tf.strings.split` function.*
- Create a function to map over the datasets that embeds each sequence of English words using the loaded embedding layer/model. Apply this function to both Dataset objects using the map method.
- Create a function to filter out dataset examples where the English sentence is greater than or equal to than 13 (embedded) tokens in length. Apply this function to both Dataset objects using the filter method.
- Create a function to map over the datasets that pads each English sequence of embeddings with some distinct padding value before the sequence, so that each sequence is length 13. Apply this function to both Dataset objects using the map method. *Hint: look at the `tf.pad` function. You can extract a Tensor shape using `tf.shape`; you might also find the `tf.math.maximum` function useful*

*function useful.*

- Batch both training and validation Datasets with a batch size of 16.
- Print the `element_spec` property for the training and validation Datasets.
- Using the Dataset `.take(1)` method, print the shape of the English data example from the training Dataset.
- Using the Dataset `.take(1)` method, print the German data example Tensor from the validation Dataset.

In [18]:

```
# Create training, test and validation splits
from random import shuffle
from sklearn.model_selection import train_test_split

english_train, english_valid, german_train, german_valid = train_test_split(english,
                                                                              padded_german_seq,
                                                                              test_size=0.2)
```

In [19]:

```
# Create a translation Dataset
eng_ger_train = tf.data.Dataset.from_tensor_slices((english_train, german_train))
eng_ger_valid = tf.data.Dataset.from_tensor_slices((english_valid, german_valid))
eng_ger_train.element_spec
```

Out[19]:

```
(TensorSpec(shape=(), dtype=tf.string, name=None),
 TensorSpec(shape=(14,), dtype=tf.int32, name=None))
```

In [20]:

```
# Split the English sentences at spaces
def split_at_spaces(eng, ger):
    return tf.strings.split(eng), ger

eng_ger_train = eng_ger_train.map(split_at_spaces)
eng_ger_valid = eng_ger_valid.map(split_at_spaces)
eng_ger_train.element_spec
```

Out[20]:

```
(TensorSpec(shape=(None,), dtype=tf.string, name=None),
 TensorSpec(shape=(14,), dtype=tf.int32, name=None))
```

In [21]:

```
# Get the embeddings for the English sentences
def get_embedding(eng, ger):
    return embedding_layer(eng), ger

eng_ger_train = eng_ger_train.map(get_embedding)
eng_ger_valid = eng_ger_valid.map(get_embedding)
eng_ger_train.element_spec
```

Out[21]:

```
(TensorSpec(shape=(None, 128), dtype=tf.float32, name=None),
 TensorSpec(shape=(14,), dtype=tf.int32, name=None))
```

In [22]:

```
# Filter out examples where the English sentence is longer than 13 tokens
max_english_sentence_len = 13

def filter_long_sentences(eng, ger):
    return tf.shape(eng)[0] <= max_english_sentence_len

eng_ger_train = eng_ger_train.filter(filter_long_sentences)
```

```
eng_ger_train = eng_ger_train.filter(filter_long_sentences)
eng_ger_valid = eng_ger_valid.filter(filter_long_sentences)
eng_ger_train.element_spec
```

Out[22]:

```
(TensorSpec(shape=(None, 128), dtype=tf.float32, name=None),
 TensorSpec(shape=(14,), dtype=tf.int32, name=None))
```

In [23]:

```
# Pad the English sentences
def pad_embedding_sequences(eng, ger):
    length_emb = tf.shape(eng)[0]
    padding = [[tf.math.maximum(0, max_english_sentence_len - length_emb), 0], [0, 0]]
    return tf.pad(eng, padding, 'CONSTANT', constant_values=-1000), ger

eng_ger_train = eng_ger_train.map(pad_embedding_sequences)
eng_ger_valid = eng_ger_valid.map(pad_embedding_sequences)
eng_ger_train.element_spec
```

Out[23]:

```
(TensorSpec(shape=(None, 128), dtype=tf.float32, name=None),
 TensorSpec(shape=(14,), dtype=tf.int32, name=None))
```

In [24]:

```
# Batch the Datasets
eng_ger_train = eng_ger_train.batch(16)
eng_ger_valid = eng_ger_valid.batch(16)
```

In [25]:

```
# Print the final training Dataset element_spec
eng_ger_train.element_spec
```

Out[25]:

```
(TensorSpec(shape=(None, None, 128), dtype=tf.float32, name=None),
 TensorSpec(shape=(None, 14), dtype=tf.int32, name=None))
```

In [26]:

```
# Print the shape of a batch of English data examples
for eng, ger in eng_ger_train.take(1):
    print(eng.shape)
```

```
(16, 13, 128)
```

In [27]:

```
# Print a batch of German data examples
for eng, ger in eng_ger_valid.take(1):
    print(ger)
```

```
tf.Tensor(
[[ 1  763  17 381  7 2 0 0 0 0 0 0 0 0]
 [ 1  5 4038 22 12 3 2 0 0 0 0 0 0 0]
 [ 1  4 3434 3 2 0 0 0 0 0 0 0 0 0]
 [ 1 10 6 625 401 3 2 0 0 0 0 0 0]
 [ 1 11 152 6 611 3 2 0 0 0 0 0 0]
 [ 1 4 918 10 141 3 2 0 0 0 0 0 0]
 [ 1 406 53 1165 9 2 0 0 0 0 0 0]
 [ 1 202 2405 1220 24 2 2 2 2 2 2 2 2 2]])
```

```
[ 1 202 2485 1329 34 3 2 0 0 0 0 0 0 0]
[ 1 4 18 27 37 1368 4835 3 2 0 0 0 0 0]
[ 1 5 165 1735 20 54 3 2 0 0 0 0 0 0]
[ 1 5 16 2694 3 2 0 0 0 0 0 0 0 0]
[ 1 4 39 10 124 220 3 2 0 0 0 0 0 0]
[ 1 4 61 119 3 2 0 0 0 0 0 0 0 0]
[ 1 4 18 40 1593 3 2 0 0 0 0 0 0 0]
[ 1 14 5568 125 3 2 0 0 0 0 0 0 0 0]
[ 1 13 246 342 92 3 2 0 0 0 0 0 0 0]], shape=(16, 14), dtype=int32)
```

### 3. Create the custom layer

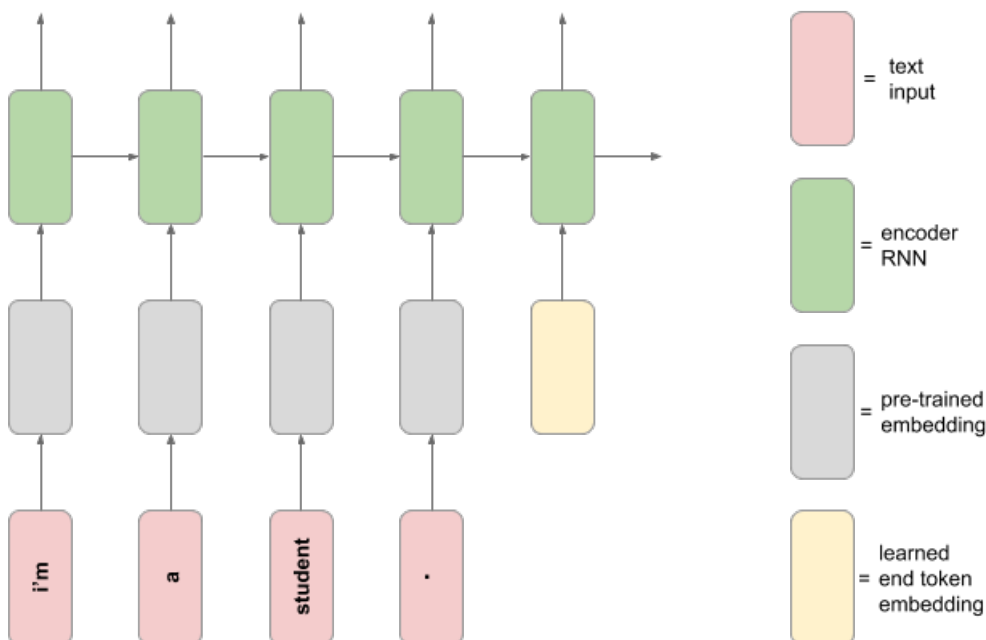
You will now create a custom layer to add the learned end token embedding to the encoder model:

In [28]:

```
# Run this cell to download and view a schematic diagram for the encoder model

!wget -q -O neural_translation_model.png --no-check-certificate "https://docs.google.com/uc?export=download&id=1JrtNOzUJDaOWrK4C-xv-4wUuZaI12sQI"
Image("neural_translation_model.png")
```

Out[28]:



You should now build the custom layer.

- Using layer subclassing, create a custom layer that takes a batch of English data examples from one of the Datasets, and adds a learned embedded 'end' token to the end of each sequence.
- This layer should create a TensorFlow Variable (that will be learned during training) that is 128-dimensional (the size of the embedding space). *Hint: you may find it helpful in the call method to use the `tf.tile` function to replicate the end token embedding across every element in the batch.*
- Using the Dataset `.take(1)` method, extract a batch of English data examples from the training Dataset and print the shape. Test the custom layer by calling the layer on the English data batch Tensor and print the resulting Tensor shape (the layer should increase the sequence length by one).

In [29]:

```
# Create the custom layer to add an end token embedding
class AddEndToken(tf.keras.layers.Layer):
    def __init__(self, **kwargs):
        super(AddEndToken, self).__init__(**kwargs)

    def build(self, input_shape):
        embedding_size = input_shape[-1]
```



```

        self.end_token = self.add_weight(shape=(1, embedding_size),
                                         initializer='random_normal',
                                         name='english_end_token')

    def call(self, inputs):
        batch_size = tf.shape(inputs)[0]
        token = tf.tile(tf.expand_dims(self.end_token, 0), [batch_size, 1, 1])
        return tf.concat([inputs, token], 1)

```

In [30]:

```

# Test the layer
add_end_token = AddEndToken()

for eng, ger in eng_ger_train.take(1):
    print(eng.shape)
    print(add_end_token(eng).shape)

```

```

(16, 13, 128)
(16, 14, 128)

```

In [30]:

## 4. Build the encoder network

The encoder network follows the schematic diagram above. You should now build the RNN encoder model.

- Using the functional API, build the encoder network according to the following spec:
  - The model will take a batch of sequences of embedded English words as input, as given by the Dataset objects.
  - The next layer in the encoder will be the custom layer you created previously, to add a learned end token embedding to the end of the English sequence.
  - This is followed by a Masking layer, with the `mask_value` set to the distinct padding value you used when you padded the English sequences with the Dataset preprocessing above.
  - The final layer is an LSTM layer with 512 units, which also returns the hidden and cell states.
  - The encoder is a multi-output model. There should be two output Tensors of this model: the hidden state and cell states of the LSTM layer. The output of the LSTM layer is unused.
- Using the Dataset `.take(1)` method, extract a batch of English data examples from the training Dataset and test the encoder model by calling it on the English data Tensor, and print the shape of the resulting Tensor outputs.
- Print the model summary for the encoder network.

In [31]:

```

# Build the encoder
from tensorflow.keras.models import Model
from tensorflow.keras.layers import Input, LSTM, Masking

def get_encoder():
    encoder_input = Input(shape=(None, 128))
    h = AddEndToken()(encoder_input)
    h = Masking(mask_value=-1000)(h)
    enc_output, enc_hidden, enc_cell = LSTM(512, return_state=True)(h)

    return Model(inputs=encoder_input, outputs=[enc_hidden, enc_cell])

encoder = get_encoder()

```

In [32]:

```

# Test the encoder
for eng, ger in eng_ger_train.take(1):
    print([t.shape for t in encoder(eng)])

```

```
[TensorShape([16, 512]), TensorShape([16, 512])]
```

```
In [33]:
```

```
# Print the encoder summary
encoder.summary()
```

```
Model: "functional_1"
```

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[ (None, None, 128) ]	0
add_end_token_1 (AddEndToken)	(None, None, 128)	128
masking (Masking)	(None, None, 128)	0
lstm (LSTM)	[ (None, 512), (None, 512) ]	1312768

Total params: 1,312,896  
Trainable params: 1,312,896  
Non-trainable params: 0

## 5. Build the decoder network

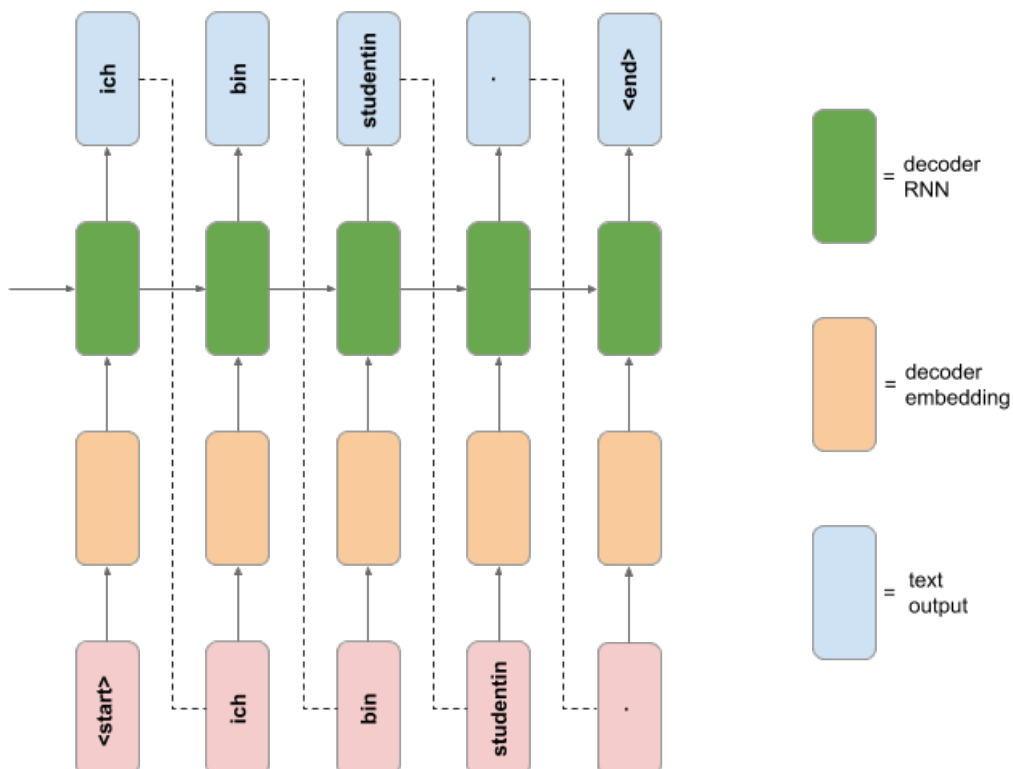
The decoder network follows the schematic diagram below.

```
In [34]:
```

```
# Run this cell to download and view a schematic diagram for the decoder model
```

```
!wget -q -O neural_translation_model.png --no-check-certificate "https://docs.google.com/uc?export=download&id=1DTeaXD8tA8RjKpVrB2mr9csSBOY4LQiW"  
Image("neural_translation_model.png")
```

```
Out[34]:
```



You should now build the RNN decoder model.

- Using Model subclassing, build the decoder network according to the following spec:
  - The initializer should create the following layers:
    - An Embedding layer with vocabulary size set to the number of unique German tokens, embedding dimension 128, and set to mask zero values in the input.
    - An LSTM layer with 512 units, that returns its hidden and cell states, and also returns sequences.
    - A Dense layer with number of units equal to the number of unique German tokens, and no activation function.
  - The call method should include the usual `inputs` argument, as well as the additional keyword arguments `hidden_state` and `cell_state`. The default value for these keyword arguments should be `None`.
  - The call method should pass the inputs through the Embedding layer, and then through the LSTM layer. If the `hidden_state` and `cell_state` arguments are provided, these should be used for the initial state of the LSTM layer. *Hint: use the `initial_state` keyword argument when calling the LSTM layer on its input.*
  - The call method should pass the LSTM output sequence through the Dense layer, and return the resulting Tensor, along with the hidden and cell states of the LSTM layer.
- Using the Dataset `.take(1)` method, extract a batch of English and German data examples from the training Dataset. Test the decoder model by first calling the encoder model on the English data Tensor to get the hidden and cell states, and then call the decoder model on the German data Tensor and hidden and cell states, and print the shape of the resulting decoder Tensor outputs.
- Print the model summary for the decoder network.

In [35]:

```
# Build the decoder
from tensorflow.keras.layers import Dense, Embedding

class Decoder(Model):
    def __init__(self, **kwargs):
        super(Decoder, self).__init__(**kwargs)
        self.embedding = Embedding(num_german_tokens, 128, mask_zero=True)
        self.lstm = LSTM(512, return_state=True, return_sequences=True)
        self.dense = Dense(num_german_tokens)

    def call(self, inputs, hidden_state=None, cell_state=None, training=True):
        if hidden_state is None:
            assert cell_state is None
            h = self.embedding(inputs)
            o, h, c = self.lstm(h)
        else:
            h = self.embedding(inputs)
            o, h, c = self.lstm(h, initial_state=[hidden_state, cell_state])
        return self.dense(o), h, c

decoder = Decoder()
```

In [36]:

```
# Test the decoder
for eng, ger in eng_ger_train.take(1):
    hidden, cell = encoder(eng)
    print([t.shape for t in decoder(ger, hidden_state=hidden, cell_state=cell)])
```

```
[TensorShape([16, 14, 5744]), TensorShape([16, 512]), TensorShape([16, 512])]
```

In [37]:

```
# Print the decoder summary
decoder.summary()
```

Model: "decoder"

Layer (type)	Output Shape	Param #
embedding (Embedding)	multiple	735232
lstm_1 (LSTM)	multiple	1312768

dense (Dense)	multiple	2946672
=====		
Total params: 4,994,672		
Trainable params: 4,994,672		
Non-trainable params: 0		

## 6. Make a custom training loop

You should now write a custom training loop to train your custom neural translation model.

- Define a function that takes a Tensor batch of German data (as extracted from the training Dataset), and returns a tuple containing German inputs and outputs for the decoder model (refer to schematic diagram above).
- Define a function that computes the forward and backward pass for your translation model. This function should take an English input, German input and German output as arguments, and should do the following:
  - Pass the English input into the encoder, to get the hidden and cell states of the encoder LSTM.
  - These hidden and cell states are then passed into the decoder, along with the German inputs, which returns a sequence of outputs (the hidden and cell state outputs of the decoder LSTM are unused in this function).
  - The loss should then be computed between the decoder outputs and the German output function argument.
  - The function returns the loss and gradients with respect to the encoder and decoder's trainable variables.
  - Decorate the function with `@tf.function`
- Define and run a custom training loop for a number of epochs (for you to choose) that does the following:
  - Iterates through the training dataset, and creates decoder inputs and outputs from the German sequences.
  - Updates the parameters of the translation model using the gradients of the function above and an optimizer object.
  - Every epoch, compute the validation loss on a number of batches from the validation and save the epoch training and validation losses.
- Plot the learning curves for loss vs epoch for both training and validation sets.

*Hint: This model is computationally demanding to train. The quality of the model or length of training is not a factor in the grading rubric. However, to obtain a better model we recommend using the GPU accelerator hardware on Colab.*

In [38]:

```
# Define the loss function and optimizer
loss_fn = tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True)
optimizer = tf.keras.optimizers.Adam()
```

In [39]:

```
# Function to create German tokens inputs and outputs
def get_german_inputs_outputs(ger_in):
    return ger_in[:, :-1], ger_in[:, 1:]
```

In [40]:

```
# Define the forward and backward pass
@tf.function
def get_loss_and_grads(eng_in, ger_in, ger_out):
    with tf.GradientTape() as tape:
        h, c = encoder(eng_in)
        outputs, _1, _2 = decoder(ger_in, hidden_state=h, cell_state=c)
        current_loss = loss_fn(ger_out, outputs)
        grads = tape.gradient(current_loss,
                               encoder.trainable_variables + decoder.trainable_variables)
    return current_loss, grads
```

In [41]:

```
# For evaluation
@tf.function
def get_loss(eng_in, ger_in, ger_out):
    h, c = encoder(eng_in)
    outputs, _1, _2 = decoder(ger_in, hidden_state=h, cell_state=c)
    return loss_fn(ger_out, outputs)
```

In [42]:

```
# Run the custom training loop
num_epochs = 8
num_valid_steps = 100

epoch_history = {
    'epoch': [],
    'loss': [],
    'val_loss': []
}
batch_history = {
    'iteration': [],
    'loss': []
}

for epoch in range(num_epochs):
    epoch_loss = tf.keras.metrics.Mean()

    for iteration, (eng, ger) in enumerate(eng_ger_train):
        ger_inputs, ger_outputs = get_german_inputs_outputs(ger)
        loss, grads = get_loss_and_grads(eng, ger_inputs, ger_outputs)
        epoch_loss(loss)
        print("Iteration {}, loss: {}".format(iteration, loss))
        if iteration % 10 == 0:
            batch_history['iteration'].append(iteration)
            batch_history['loss'].append(loss)
        optimizer.apply_gradients(zip(grads, encoder.trainable_variables + decoder.trainable_variables))

    print("End of epoch {}, loss: {}".format(epoch, epoch_loss.result() + 1))

    # Validation
    for (eng, ger) in eng_ger_valid.take(num_valid_steps):
        valid_loss = tf.keras.metrics.Mean()
        ger_inputs, ger_outputs = get_german_inputs_outputs(ger)
        valid_loss(get_loss(eng, ger_inputs, ger_outputs))
    print("Validation loss: {}".format(valid_loss.result()))

    epoch_history['epoch'].append(epoch)
    epoch_history['loss'].append(epoch_loss.result())
    epoch_history['val_loss'].append(valid_loss.result())
```

```
Iteration 146, loss: 2.8058855533599854
Iteration 147, loss: 2.7952470779418945
Iteration 148, loss: 2.9884378910064697
Iteration 149, loss: 2.950084686279297
Iteration 150, loss: 2.741946220397949
Iteration 151, loss: 2.887845754623413
Iteration 152, loss: 2.7155299186706543
Iteration 153, loss: 2.885842800140381
Iteration 154, loss: 2.812913656234741
Iteration 155, loss: 2.878556251525879
Iteration 156, loss: 2.9246485233306885
Iteration 157, loss: 2.775843620300293
Iteration 158, loss: 2.751404047012329
Iteration 159, loss: 2.5388360023498535
Iteration 160, loss: 2.8655099868774414
Iteration 161, loss: 2.964594841003418
Iteration 162, loss: 2.7003533840179443
Iteration 163, loss: 2.8629777431488037
Iteration 164, loss: 2.6988489627838135
Iteration 165, loss: 2.719658851623535
Iteration 166, loss: 2.761998414993286
Iteration 167, loss: 2.8764612674713135
Iteration 168, loss: 2.8896565437316895
Iteration 169, loss: 2.9137344360351562
Iteration 170, loss: 2.7605552673339844
Iteration 171, loss: 3.0624351501464844
Iteration 172, loss: 2.89264178276062
Iteration 173, loss: 2.909329652786255
Iteration 174, loss: 2.7756612300872803
Iteration 175, loss: 2.8397061824798584
Iteration 176, loss: 2.738278865814209
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Iteration 989, loss: 2.1785311698913574  
Iteration 990, loss: 2.206315517425537  
Iteration 991, loss: 2.245161294937134  
Iteration 992, loss: 2.2169077396392822  
Iteration 993, loss: 2.1678836345672607  
Iteration 994, loss: 2.1389639377593994  
Iteration 995, loss: 2.180595636367798  
Iteration 996, loss: 2.12227463722229  
Iteration 997, loss: 2.4014694690704346  
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Iteration 999, loss: 2.094130277633667  
End of epoch 3, loss: 3.567085027694702  
Validation loss: 2.443629264831543  
Iteration 0, loss: 2.254446268081665  
Iteration 1, loss: 2.1170618534088135  
Iteration 2, loss: 2.1007015705108643  
Iteration 3, loss: 2.181905746459961  
Iteration 4, loss: 2.156905174255371  
Iteration 5, loss: 2.189368724822998  
Iteration 6, loss: 2.224612236022949  
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Iteration 8, loss: 2.167840003967285  
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Iteration 10, loss: 2.1681530475616455  
Iteration 11, loss: 2.202368974685669  
Iteration 12, loss: 2.235257625579834  
Iteration 13, loss: 2.1902976036071777  
Iteration 14, loss: 2.2640342712402344  
Iteration 15, loss: 2.219418525695801  
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Iteration 17, loss: 2.1859631538391113  
Iteration 18, loss: 2.2959864139556885  
Iteration 19, loss: 2.2147140502929688  
Iteration 20, loss: 2.2737345695495605  
Iteration 21, loss: 2.2330756187438965  
Iteration 22, loss: 2.2220215767424216



Iteration 22, loss: 2.2220215797424510  
Iteration 23, loss: 2.0712525844573975  
Iteration 24, loss: 2.0518171787261963  
Iteration 25, loss: 2.241594076156616  
Iteration 26, loss: 2.292088508605957  
Iteration 27, loss: 2.2811620235443115  
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Iteration 33, loss: 2.3499794006347656  
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Iteration 98, loss: 2.0973660945892334  
Iteration 99, loss: 2.1000000000000000

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Iteration 888, loss: 1.4941961765289307  
Iteration 889, loss: 1.569062352180481  
Iteration 890, loss: 1.5733530521392822  
Iteration 891, loss: 1.621049404144287  
Iteration 892, loss: 1.5249228477478027  
Iteration 893, loss: 1.6075317859649658  
Iteration 894, loss: 1.6103771924972534  
Iteration 895, loss: 1.528605580329895  
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Iteration 902, loss: 1.5127589702606201  
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Iteration 904, loss: 1.5577139854431152  
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Iteration 907, loss: 1.5503450632095337  
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Iteration 979, loss: 1.4829108715057373  
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Iteration 984, loss: 1.476143479347229  
Iteration 985, loss: 1.546494960784912  
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Iteration 987, loss: 1.5307673215866089  
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End of epoch 4, loss: 2.8222713470458984  
Validation loss: 1.8756128549575806  
Iteration 0, loss: 1.5596344470977783  
Iteration 1, loss: 1.4363118410110474  
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Iteration 3, loss: 1.4420262575149536  
Iteration 4, loss: 1.482550859451294  
Iteration 5, loss: 1.5007649660110474  
Iteration 6, loss: 1.5038306713104248  
Iteration 7, loss: 1.4295140504837036  
Iteration 8, loss: 1.4806791543960571  
Iteration 9, loss: 1.4236963987350464  
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Iteration 13, loss: 1.4739562273025513  
Iteration 14, loss: 1.5672872066497803  
Iteration 15, loss: 1.5133156776428223  
Iteration 16, loss: 1.5366578102111816  
Iteration 17, loss: 1.5183762311935425  
Iteration 18, loss: 1.599089503288269  
Iteration 19, loss: 1.5120564699172974  
Iteration 20, loss: 1.548127293586731

Iteration 21, loss: 1.5427354574203491  
Iteration 22, loss: 1.5443406105041504  
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Iteration 24, loss: 1.3805819749832153  
Iteration 25, loss: 1.537406086921692  
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Iteration 785, loss: 0.9759792685508728  
Iteration 786, loss: 1.0039387941360474  
Iteration 787, loss: 1.0641041994094849  
Iteration 788, loss: 1.059782862663269  
Iteration 789, loss: 1.1464123725891113

Iteration 790, loss: 1.033055305480957  
Iteration 791, loss: 1.0516890287399292  
Iteration 792, loss: 1.0324759483337402  
Iteration 793, loss: 1.031847357749939  
Iteration 794, loss: 0.96539306640625  
Iteration 795, loss: 1.037819743156433  
Iteration 796, loss: 1.1315970420837402  
Iteration 797, loss: 0.9769617915153503  
Iteration 798, loss: 1.0127092599868774  
Iteration 799, loss: 1.0224846601486206  
Iteration 800, loss: 1.000767707824707  
Iteration 801, loss: 0.9691574573516846  
Iteration 802, loss: 1.1102871894836426  
Iteration 803, loss: 0.9069645404815674  
Iteration 804, loss: 1.0200625658035278  
Iteration 805, loss: 1.0427144765853882  
Iteration 806, loss: 0.9803057312965393  
Iteration 807, loss: 0.9992090463638306  
Iteration 808, loss: 1.0651081800460815  
Iteration 809, loss: 1.0645346641540527  
Iteration 810, loss: 1.0867940187454224  
Iteration 811, loss: 1.0010786056518555  
Iteration 812, loss: 1.0726287364959717  
Iteration 813, loss: 0.9574878215789795  
Iteration 814, loss: 1.0588793754577637  
Iteration 815, loss: 1.0765992403030396  
Iteration 816, loss: 0.9646514058113098  
Iteration 817, loss: 1.0525171756744385  
Iteration 818, loss: 1.0419856309890747  
Iteration 819, loss: 1.071085810661316  
Iteration 820, loss: 1.0462937355041504  
Iteration 821, loss: 1.136576533317566  
Iteration 822, loss: 1.1033360958099365  
Iteration 823, loss: 0.9792537093162537  
Iteration 824, loss: 1.0599274635314941  
Iteration 825, loss: 0.9919744729995728  
Iteration 826, loss: 1.0399147272109985  
Iteration 827, loss: 1.050578236579895  
Iteration 828, loss: 0.9962809681892395  
Iteration 829, loss: 1.046187162399292  
Iteration 830, loss: 0.9891966581344604  
Iteration 831, loss: 1.1353589296340942  
Iteration 832, loss: 0.9546563625335693  
Iteration 833, loss: 0.9818232655525208  
Iteration 834, loss: 0.985697865486145  
Iteration 835, loss: 1.1572197675704956  
Iteration 836, loss: 0.983454704284668  
Iteration 837, loss: 1.0309392213821411  
Iteration 838, loss: 0.9922838807106018  
Iteration 839, loss: 0.9989707469940186  
Iteration 840, loss: 1.0082650184631348  
Iteration 841, loss: 1.0596593618392944  
Iteration 842, loss: 1.004831075668335  
Iteration 843, loss: 0.9747257232666016  
Iteration 844, loss: 1.002037763595581  
Iteration 845, loss: 0.9866189956665039  
Iteration 846, loss: 0.9943544864654541  
Iteration 847, loss: 1.0333805084228516  
Iteration 848, loss: 1.137121319770813  
Iteration 849, loss: 1.1088879108428955  
Iteration 850, loss: 0.9335642457008362  
Iteration 851, loss: 1.0199995040893555  
Iteration 852, loss: 1.0189582109451294  
Iteration 853, loss: 1.0619728565216064  
Iteration 854, loss: 0.984007716178894  
Iteration 855, loss: 0.9985535740852356  
Iteration 856, loss: 0.9908421039581299  
Iteration 857, loss: 1.1250654458999634  
Iteration 858, loss: 1.0674293041229248  
Iteration 859, loss: 1.0037264823913574  
Iteration 860, loss: 1.0948410034179688  
Iteration 861, loss: 1.0418384075164795  
Iteration 862, loss: 0.9903180003166199  
Iteration 863, loss: 1.0363224744796753  
Iteration 864, loss: 0.9704148173332214  
Iteration 865, loss: 1.0656566619873047  
Iteration 866, loss: 1.0128123760223389



Iteration 867, loss: 1.0257151126861572  
Iteration 868, loss: 1.0159071683883667  
Iteration 869, loss: 0.9614028930664062  
Iteration 870, loss: 1.0109275579452515  
Iteration 871, loss: 1.0173031091690063  
Iteration 872, loss: 0.9866231679916382  
Iteration 873, loss: 0.9947823882102966  
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Iteration 875, loss: 0.9715680480003357  
Iteration 876, loss: 0.9265117049217224  
Iteration 877, loss: 0.9150563478469849  
Iteration 878, loss: 1.1126152276992798  
Iteration 879, loss: 0.9948216676712036  
Iteration 880, loss: 0.9694774150848389  
Iteration 881, loss: 0.9983922243118286  
Iteration 882, loss: 0.9850447773933411  
Iteration 883, loss: 0.9667062163352966  
Iteration 884, loss: 1.014098882675171  
Iteration 885, loss: 1.007206916809082  
Iteration 886, loss: 1.0747041702270508  
Iteration 887, loss: 1.0025367736816406  
Iteration 888, loss: 0.937017560005188  
Iteration 889, loss: 0.9806214570999146  
Iteration 890, loss: 1.008898138999939  
Iteration 891, loss: 1.0559779405593872  
Iteration 892, loss: 0.9865044951438904  
Iteration 893, loss: 1.008971095085144  
Iteration 894, loss: 1.0506705045700073  
Iteration 895, loss: 0.9891240000724792  
Iteration 896, loss: 0.9714454412460327  
Iteration 897, loss: 0.9837590456008911  
Iteration 898, loss: 1.014839768409729  
Iteration 899, loss: 0.9376802444458008  
Iteration 900, loss: 1.0147075653076172  
Iteration 901, loss: 0.9912852048873901  
Iteration 902, loss: 0.986790657043457  
Iteration 903, loss: 0.8947153687477112  
Iteration 904, loss: 1.0114187002182007  
Iteration 905, loss: 1.0754412412643433  
Iteration 906, loss: 1.005761981010437  
Iteration 907, loss: 1.0060688257217407  
Iteration 908, loss: 1.0203347206115723  
Iteration 909, loss: 0.9941647052764893  
Iteration 910, loss: 1.0148733854293823  
Iteration 911, loss: 0.9642781019210815  
Iteration 912, loss: 0.9867672324180603  
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Iteration 914, loss: 0.8923453092575073  
Iteration 915, loss: 1.013697624206543  
Iteration 916, loss: 0.9909569621086121  
Iteration 917, loss: 0.8884525895118713  
Iteration 918, loss: 0.9814977049827576  
Iteration 919, loss: 1.006693720817566  
Iteration 920, loss: 0.9485213160514832  
Iteration 921, loss: 0.9991140961647034  
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Iteration 923, loss: 0.9836858510971069  
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Iteration 925, loss: 1.0495641231536865  
Iteration 926, loss: 0.9680442810058594  
Iteration 927, loss: 0.9582633972167969  
Iteration 928, loss: 1.007190465927124  
Iteration 929, loss: 0.83547043800354  
Iteration 930, loss: 1.0606186389923096  
Iteration 931, loss: 0.9660081267356873  
Iteration 932, loss: 0.925424337387085  
Iteration 933, loss: 0.9820288419723511  
Iteration 934, loss: 1.0433297157287598  
Iteration 935, loss: 0.9466943144798279  
Iteration 936, loss: 0.9892946481704712  
Iteration 937, loss: 0.9938830137252808  
Iteration 938, loss: 0.9421875476837158  
Iteration 939, loss: 1.0117024183273315  
Iteration 940, loss: 0.9662874937057495  
Iteration 941, loss: 0.9688053727149963  
Iteration 942, loss: 0.9305317401885986  
Iteration 943, loss: 1.0092110633850098

Iteration 944, loss: 1.0608879327774048  
Iteration 945, loss: 1.0059542655944824  
Iteration 946, loss: 0.9651960730552673  
Iteration 947, loss: 0.9539811015129089  
Iteration 948, loss: 0.9463104605674744  
Iteration 949, loss: 0.8627492189407349  
Iteration 950, loss: 0.9553599953651428  
Iteration 951, loss: 0.9486192464828491  
Iteration 952, loss: 0.9403049945831299  
Iteration 953, loss: 0.9425152540206909  
Iteration 954, loss: 0.9703810214996338  
Iteration 955, loss: 0.9836364388465881  
Iteration 956, loss: 0.9511006474494934  
Iteration 957, loss: 0.9630511999130249  
Iteration 958, loss: 0.8860827088356018  
Iteration 959, loss: 0.9539110660552979  
Iteration 960, loss: 0.9940687417984009  
Iteration 961, loss: 0.8565061092376709  
Iteration 962, loss: 0.9694800972938538  
Iteration 963, loss: 1.0459938049316406  
Iteration 964, loss: 0.9500558376312256  
Iteration 965, loss: 0.9824050068855286  
Iteration 966, loss: 1.007138967514038  
Iteration 967, loss: 0.9338066577911377  
Iteration 968, loss: 0.9045533537864685  
Iteration 969, loss: 0.9914922118186951  
Iteration 970, loss: 1.0180631875991821  
Iteration 971, loss: 0.9367100596427917  
Iteration 972, loss: 0.9499215483665466  
Iteration 973, loss: 0.9956591129302979  
Iteration 974, loss: 0.9441712498664856  
Iteration 975, loss: 0.9560216069221497  
Iteration 976, loss: 0.9894440770149231  
Iteration 977, loss: 0.9818942546844482  
Iteration 978, loss: 0.9236741662025452  
Iteration 979, loss: 0.9459839463233948  
Iteration 980, loss: 0.864730715751648  
Iteration 981, loss: 0.9284086227416992  
Iteration 982, loss: 0.8822657465934753  
Iteration 983, loss: 0.8922047019004822  
Iteration 984, loss: 0.9243815541267395  
Iteration 985, loss: 0.9781865477561951  
Iteration 986, loss: 0.9680315256118774  
Iteration 987, loss: 0.9671944975852966  
Iteration 988, loss: 0.9228289723396301  
Iteration 989, loss: 0.9196102619171143  
Iteration 990, loss: 0.9619083404541016  
Iteration 991, loss: 1.0114213228225708  
Iteration 992, loss: 0.9738110303878784  
Iteration 993, loss: 0.9120689630508423  
Iteration 994, loss: 0.90351402759552  
Iteration 995, loss: 0.9014826416969299  
Iteration 996, loss: 0.9408146142959595  
Iteration 997, loss: 1.1411586999893188  
Iteration 998, loss: 0.9159939289093018  
Iteration 999, loss: 0.8669404983520508  
End of epoch 5, loss: 2.1984479427337646  
Validation loss: 1.4300923347473145  
Iteration 0, loss: 1.0218559503555298  
Iteration 1, loss: 0.9072954654693604  
Iteration 2, loss: 0.889885663986206  
Iteration 3, loss: 0.8973697423934937  
Iteration 4, loss: 0.962757408618927  
Iteration 5, loss: 1.003585696220398  
Iteration 6, loss: 0.9648170471191406  
Iteration 7, loss: 0.8942257761955261  
Iteration 8, loss: 0.9265689253807068  
Iteration 9, loss: 0.8994598388671875  
Iteration 10, loss: 0.9685699939727783  
Iteration 11, loss: 0.9619287848472595  
Iteration 12, loss: 1.0070717334747314  
Iteration 13, loss: 0.9200350046157837  
Iteration 14, loss: 1.0055615901947021  
Iteration 15, loss: 0.9440730810165405  
Iteration 16, loss: 0.9829254746437073  
Iteration 17, loss: 0.9933240413665771  
Iteration 18, loss: 1.0272390842437744

Iteration 19, loss: 0.9790473580360413  
Iteration 20, loss: 1.024342656135559  
Iteration 21, loss: 1.0100280046463013  
Iteration 22, loss: 0.9830559492111206  
Iteration 23, loss: 0.8786798119544983  
Iteration 24, loss: 0.8832809925079346  
Iteration 25, loss: 1.0188575983047485  
Iteration 26, loss: 1.024062156677246  
Iteration 27, loss: 1.0134778022766113  
Iteration 28, loss: 0.9288766384124756  
Iteration 29, loss: 0.9520953297615051  
Iteration 30, loss: 0.9542974829673767  
Iteration 31, loss: 0.9148282408714294  
Iteration 32, loss: 0.9214850068092346  
Iteration 33, loss: 1.0167758464813232  
Iteration 34, loss: 0.9834414124488831  
Iteration 35, loss: 0.9154990315437317  
Iteration 36, loss: 0.897251546382904  
Iteration 37, loss: 0.8554385900497437  
Iteration 38, loss: 0.9193137288093567  
Iteration 39, loss: 0.9528558254241943  
Iteration 40, loss: 0.9410148859024048  
Iteration 41, loss: 1.0019587278366089  
Iteration 42, loss: 0.8760501146316528  
Iteration 43, loss: 0.9111246466636658  
Iteration 44, loss: 0.8843404054641724  
Iteration 45, loss: 1.0232337713241577  
Iteration 46, loss: 0.9777405261993408  
Iteration 47, loss: 0.9098526239395142  
Iteration 48, loss: 0.9357474446296692  
Iteration 49, loss: 0.9551281332969666  
Iteration 50, loss: 0.9651819467544556  
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Iteration 52, loss: 0.921195924282074  
Iteration 53, loss: 0.9435252547264099  
Iteration 54, loss: 0.8728234767913818  
Iteration 55, loss: 1.0499809980392456  
Iteration 56, loss: 0.8203508853912354  
Iteration 57, loss: 0.9703998565673828  
Iteration 58, loss: 1.0189473628997803  
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Iteration 60, loss: 0.8718922138214111  
Iteration 61, loss: 1.0282715559005737  
Iteration 62, loss: 0.9866800904273987  
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Iteration 64, loss: 0.8812845945358276  
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Iteration 68, loss: 0.9256178140640259  
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Iteration 95, loss: 0.9320775270462036

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Iteration 110, loss: 0.9594384431838989  
Iteration 111, loss: 0.9755893349647522  
Iteration 112, loss: 0.849632203578949  
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Iteration 115, loss: 0.9198826551437378  
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Iteration 171, loss: 1.030239224433899  
Iteration 172, loss: 0.9046831130981445

Iteration 173, loss: 0.8558511734008789  
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Iteration 224, loss: 0.9158393144607544  
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Iteration 709, loss: 0.7262438535690308  
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Iteration 711, loss: 0.6600474171630480

Iteration 711, loss: 0.6920474171638489  
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Iteration 713, loss: 0.663760244846344  
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Iteration 786, loss: 0.6435518264770508  
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Iteration 788, loss: 0.682115117931366  
Iteration 789, loss: 0.7402273416519165  
Iteration 790, loss: 0.6529261469841003  
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Iteration 942, loss: 0.5733956098556519  
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End of epoch 6, loss: 1.7589681148529053  
Validation loss: 1.1912322044372559  
Iteration 0, loss: 0.6665000915527344  
Iteration 1, loss: 0.5462044477462769  
Iteration 2, loss: 0.5681434273719788  
Iteration 3, loss: 0.5382514595985413  
Iteration 4, loss: 0.6469284892082214  
Iteration 5, loss: 0.6145490407943726  
Iteration 6, loss: 0.5895761251449585  
Iteration 7, loss: 0.5469764471054077  
Iteration 8, loss: 0.5905635356903076  
Iteration 9, loss: 0.565906822681427  
Iteration 10, loss: 0.632039487361908  
Iteration 11, loss: 0.6332642436027527  
Iteration 12, loss: 0.669865071773529  
Iteration 13, loss: 0.5474310517311096  
Iteration 14, loss: 0.6287732720375061  
Iteration 15, loss: 0.6013845801353455  
Iteration 16, loss: 0.6191433668136597

Iteration 17, loss: 0.6411393880844116  
Iteration 18, loss: 0.6649078726768494  
Iteration 19, loss: 0.6173925995826721  
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Iteration 93, loss: 0.637019693851471



Iteration 94, loss: 0.5519654750823975  
Iteration 95, loss: 0.57383131980896  
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Iteration 103, loss: 0.4942598044872284  
Iteration 104, loss: 0.6977672576904297  
Iteration 105, loss: 0.6209660172462463  
Iteration 106, loss: 0.6033396124839783  
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Iteration 997, loss: 0.55936598777771
Iteration 998, loss: 0.35163599252700806
Iteration 999, loss: 0.35530751943588257
End of epoch 7, loss: 1.4969019889831543
Validation loss: 1.0703672170639038

```

In [43]:

```

# Save the complete models

decoder.save('models/decoder')
encoder.save('models/encoder')

# decoder = tf.keras.models.load_model('models/decoder')
# encoder = tf.keras.models.load_model('models/encoder')

```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/training/tracking/tracking.py:111: Model state updates (from tensorflow.python.keras engine training) is deprecated and will

king.py:111: Model.state\_updates (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.

Instructions for updating:

This property should not be used in TensorFlow 2.0, as updates are applied automatically.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/training/tracking/tracking.py:111: Model.state\_updates (from tensorflow.python.keras.engine.training) is deprecated and will be removed in a future version.

Instructions for updating:

This property should not be used in TensorFlow 2.0, as updates are applied automatically.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/training/tracking/tracking.py:111: Layer.updates (from tensorflow.python.keras.engine.base\_layer) is deprecated and will be removed in a future version.

Instructions for updating:

This property should not be used in TensorFlow 2.0, as updates are applied automatically.

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/training/tracking/tracking.py:111: Layer.updates (from tensorflow.python.keras.engine.base\_layer) is deprecated and will be removed in a future version.

Instructions for updating:

This property should not be used in TensorFlow 2.0, as updates are applied automatically.

INFO:tensorflow:Assets written to: models/decoder/assets

INFO:tensorflow:Assets written to: models/decoder/assets

INFO:tensorflow:Assets written to: models/encoder/assets

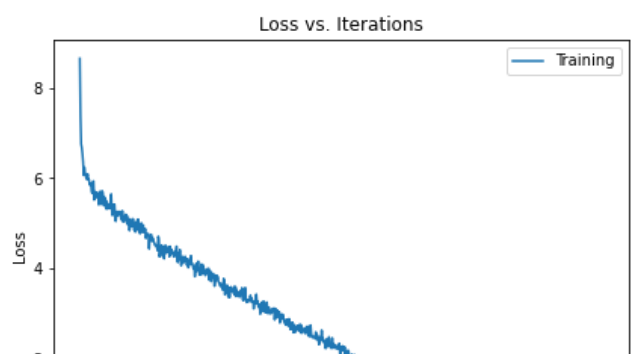
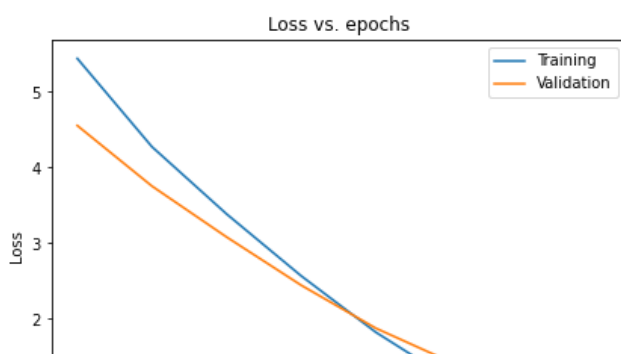
INFO:tensorflow:Assets written to: models/encoder/assets

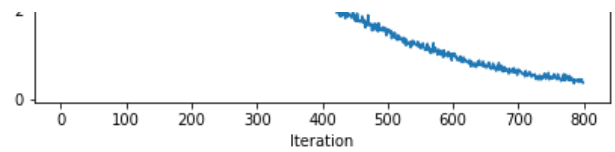
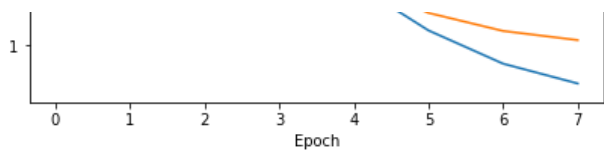
In [44]:

```
# Plot the learning curves
import matplotlib.pyplot as plt
plt.figure(figsize=(15,5))

plt.subplot(121)
plt.plot(epoch_history['loss'])
plt.plot(epoch_history['val_loss'])
plt.title('Loss vs. epochs')
plt.ylabel('Loss')
plt.xlabel('Epoch')
plt.legend(['Training', 'Validation'], loc='upper right')

plt.subplot(122)
plt.plot(batch_history['loss'])
plt.title('Loss vs. Iterations')
plt.ylabel('Loss')
plt.xlabel('Iteration')
plt.legend(['Training'], loc='upper right')
plt.show()
```





## 7. Use the model to translate

Now it's time to put your model into practice! You should run your translation for five randomly sampled English sentences from the dataset. For each sentence, the process is as follows:

- Preprocess and embed the English sentence according to the model requirements.
- Pass the embedded sentence through the encoder to get the encoder hidden and cell states.
- Starting with the special "<start>" token, use this token and the final encoder hidden and cell states to get the one-step prediction from the decoder, as well as the decoder's updated hidden and cell states.
- Create a loop to get the next step prediction and updated hidden and cell states from the decoder, using the most recent hidden and cell states. Terminate the loop when the "<end>" token is emitted, or when the sentence has reached a maximum length.
- Decode the output token sequence into German text and print the English text and the model's German translation.

In [45]:

```
# Sample a random English sentence to translate

for _ in range(5):
    english_sentence = english[np.random.choice(len(english))]
    print("English input: \t\t", re.sub(r" ([?.! ,])", r"\1 ", english_sentence))

    max_translation_length = 10

    german_translation = ["<start>"]
    ger_token = german_tokenizer.texts_to_sequences(german_translation)
    ger_token = np.array(ger_token)

    embedded_sentence = embedding_layer(tf.constant(english_sentence.split(' ')))

    hidden, cell = encoder(embedded_sentence[np.newaxis, ...])

    while True:
        decoder_output, hidden, cell = decoder(ger_token, hidden_state=hidden, cell_state=cell)
        ger_token = np.argmax(np.squeeze(decoder_output))
        if ger_token == 0:
            break
        ger_word = german_tokenizer.index_word[ger_token]
        ger_token = ger_token[np.newaxis, np.newaxis, ...]
        if ger_word == "<end>":
            break
        german_translation.append(ger_word)
        if len(german_translation) >= max_translation_length:
            break

    sentence = ' '.join(german_translation[1:])
    print("Model translation:\t", re.sub(r" ([?.! ,])", r"\1 ", sentence))
    print('')
```

English input: tom can't sing.  
Model translation: tom kann nicht zaehlen.

English input: how perceptive!  
Model translation: wie aufregend!

English input: could i ask why?  
Model translation: darf ich fragen, warum?

English input: what's it worth?  
Model translation: was ist das nun?

English input: i'll be glad to.  
Model translation: ich werde brav sein.

