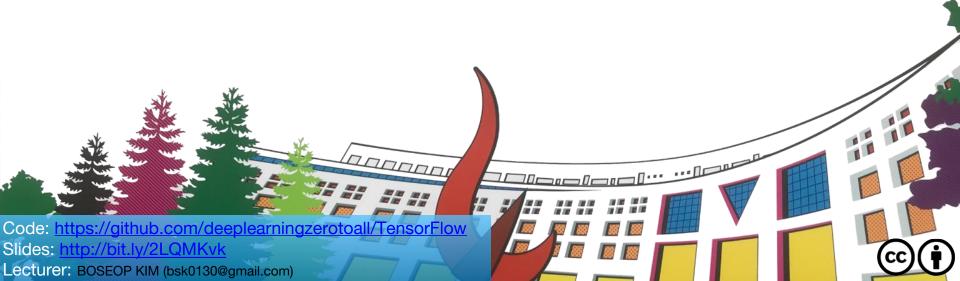
# ML/DL for Everyone Season2



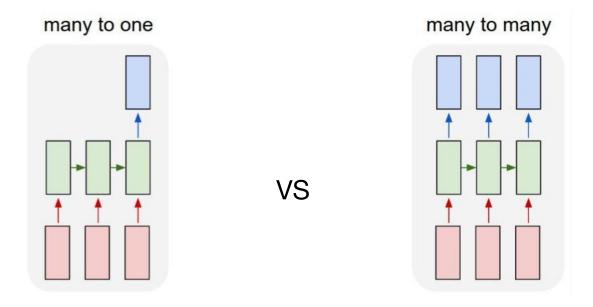
# Lab 12-3 many to many



#### many to many

- What is "many to many"?
- Example : part of speech tagging
  - Preparing dataset
  - Creating and training model
  - Checking performance

### What is "many to many"?



producing an output for final input it reads in.

producing an output for each input it reads in.

### What is "many to many"?

#### Sequence tagging

eg. part of speech tagging sequence : sentence, tokens : word

['tensorflow is very easy']

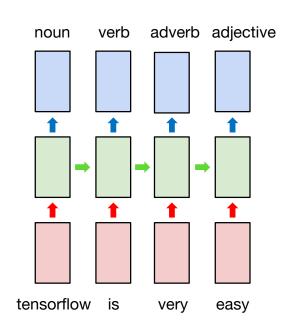
↓ Tokenization

['tensorflow', 'is', 'very', 'easy']

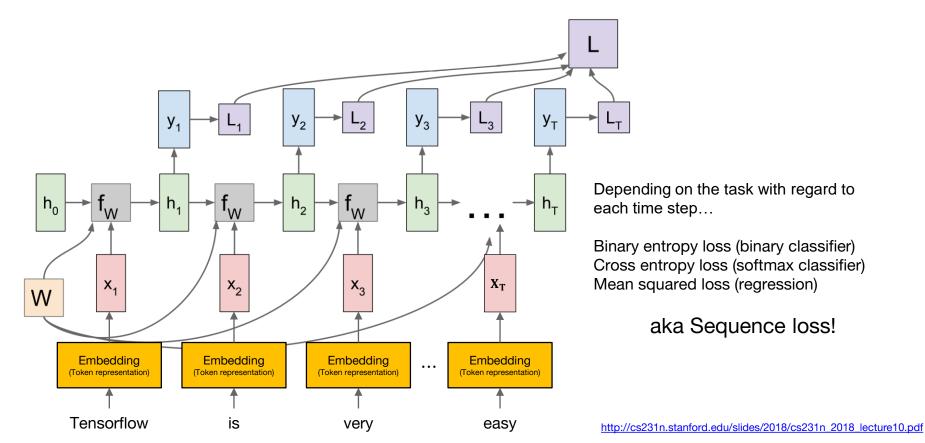
↓ Tagging

['noun', 'verb', 'adverb', 'adjective']

classification (each time step)



### What is "many to many"?



#### Preparing dataset

```
# example data
sentences = [['I', 'feel', 'hungry'],
     ['tensorflow', 'is', 'very', 'difficult'],
     ['tensorflow', 'is', 'a', 'framework', 'for', 'deep', 'learning'],
     ['tensorflow', 'is', 'very', 'fast', 'changing']]
pos = [['pronoun', 'verb', 'adjective'],
     ['noun', 'verb', 'adverb', 'adjective'],
     ['noun', 'verb', 'determiner', 'noun', 'preposition', 'adjective', 'noun'],
     ['noun', 'verb', 'adverb', 'adjective', 'verb']]
# creating a token dictionary for word
                                                                            # creating a token dictionary for part of speech
word list = sum(sentences, [])
                                                                            pos list = sum(pos, [])
word list = sorted(set(word list))
                                                                            pos list = sorted(set(pos list))
word list = ['<pad>'] + word list
                                                                            pos list = ['<pad>'] + pos list
word2idx = {word : idx for idx, word in enumerate(word list)}
                                                                            pos2idx = {pos : idx for idx, pos in enumerate(pos list)}
idx2word = {idx : word for idx, word in enumerate(word list)}
                                                                            idx2pos = {idx : pos for idx, pos in enumerate(pos list)}
                            {'<pad>': 0, 'I': 1, 'a': 2, 'changing': 3, 'deep': 4,
                                                                                                       {'<pad>': 0, 'adjective': 1, 'adverb': 2, 'determiner': 3,
print(word2idx)
                                                                            print(pos2idx)
                            'difficult': 5, 'fast': 6, 'feel': 7, 'for': 8, 'framework':
                                                                                                       'noun': 4, 'preposition': 5, 'pronoun': 6, 'verb': 7}
                           9, 'hungry': 10, 'is': 11, 'learning': 12, 'tensorflow': 13,
                                                                                                       {0: '<pad>', 1: 'adjective', 2: 'adverb', 3: 'determiner',
print(idx2word)
                                                                            print(idx2pos)
                            'verv': 14}
                                                                                                       4: 'noun', 5: 'preposition', 6: 'pronoun', 7: 'verb'}
print(len(idx2word))
                            {0: '<pad>', 1: 'I', 2: 'a', 3: 'changing', 4: 'deep', 5:
                                                                            print(len(pos2idx))
                            'difficult', 6: 'fast', 7: 'feel', 8: 'for', 9: 'framework',
                            10: 'hungry', 11: 'is', 12: 'learning', 13: 'tensorflow', 14:
                            'very'}
                            15
```

#### Preparing dataset

```
# converting sequence of tokens to sequence of indices
\max sequence = 10
x data = list(map(lambda sentence : [word2idx.get(token) for token in sentence], sentences))
y data = list(map(lambda sentence : [pos2idx.get(token) for token in sentence], pos))
# padding the sequence of indices
x_data = pad_sequences(sequences = x_data, maxlen = max_sequence, padding='post')
x data mask = ((x data != 0) * 1).astype(np.float32)
x data len = list(map(lambda sentence : len(sentence), sentences))
y data = pad sequences(sequences = y data, maxlen = max sequence, padding='post')
# checking data
print(x data, x data len)
                                [13 11 2 9 8 4 12 0 0 0]
print(x data mask)
                                [13 11 14 6 3 0 0 0 0 0]] [3, 4, 7, 5]
print(y data)
                                [[1. 1. 1. 0. 0. 0. 0. 0. 0. 0.]
                                [1. 1. 1. 1. 0. 0. 0. 0. 0. 0.]
                                [1. 1. 1. 1. 1. 1. 0. 0. 0.]
                                [1. 1. 1. 1. 1. 0. 0. 0. 0. 0. ]]
                                 [[6 7 1 0 0 0 0 0 0 0]
                                 [4 7 2 1 0 0 0 0 0 0]
                                 [4 7 3 4 5 1 4 0 0 0]
                                 [4 7 2 1 7 0 0 0 0 0]]
```

https://github.com/deeplearningzerotoall/TensorFlow/blob/master/lab-12-3-many-to-many-keras-eager.ipynb

#### Creating and training model

```
# creating rnn for "many to many" sequence tagging
num classes = len(pos2idx)
hidden dim = 10
input dim = len(word2idx)
output dim = len(word2idx)
one hot = np.eye(len(word2idx))
model = Sequential()
model.add(layers.Embedding(input dim=input dim, output dim=output dim, mask zero=True,
                             trainable=False, input_length=max_sequence,
                             embeddings initializer=keras.initializers.Constant(one hot)))
model.add(layers.SimpleRNN(units=hidden dim, return sequences=True))
                                                                                  Layer (type)
                                                                                                       Output Shape.
                                                                                                                          Param #
model.add(layers.TimeDistributed(layers.Dense(units=num classes)))
model.summary()
                                                                                 embedding (Embedding)
                                                                                                       (None, 10, 15)
                                                                                                                          225
                                                                                 simple rnn (SimpleRNN)
                                                                                                       (None, 10, 10)
                                                                                                                          260
                                                                                  time distributed (TimeDistri (None, 18, 8)
                                                                                  Total params: 573
                                                                                 Trainable params: 348
                                                                                  Non-trainable params: 225
```

#### Creating and training model

```
# creating loss function
def loss fn(model, x, y, x len, max sequence):
   masking = tf.sequence mask(x len, maxlen=max sequence, dtype=tf.float32)
   valid time step = tf.cast(x len,dtype=tf.float32)
   sequence loss = tf.losses.sparse softmax cross entropy(labels=y, logits=model(x),
                                                            reduction='none') * masking
   sequence loss = tf.reduce sum(sequence loss, axis=-1) / valid time step
   sequence loss = tf.reduce mean(sequence loss)
   return sequence loss
# creating and optimizer
1r = 0.1
epochs = 30
batch size = 2
opt = tf.train.AdamOptimizer(learning_rate = lr)
# generating data pipeline
tr dataset = tf.data.Dataset.from tensor slices((x data, y data, x data len))
tr dataset = tr dataset.shuffle(buffer size=4)
tr dataset = tr dataset.batch(batch size = 2)
                                                      <BatchDataset shapes: ((?, 10), (?, 10), (?,)), types: (tf.int32, tf.int32,</pre>
                                                      tf.int32)>
print(tr dataset)
```

https://github.com/deeplearningzerotoall/TensorFlow/blob/master/lab-12-3-many-to-many-keras-eager.ipynb

#### Creating and training model

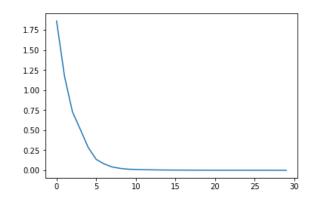
```
# trainina
tr loss hist = []
for epoch in range(epochs):
   avg tr loss = 0
   tr step = 0
   for x mb, y mb, x mb len in tr dataset:
       with tf.GradientTape() as tape:
           tr loss = loss fn(model, x=x mb, y=y mb, x len=x mb len, max sequence=max sequence)
       grads = tape.gradient(target=tr loss, sources=model.variables)
       opt.apply gradients(grads_and_vars=zip(grads, model.variables))
                                                                                           epoch: 5, tr loss: 0.284
       avg tr loss += tr loss
                                                                                           epoch : 10, tr loss : 0.015
       tr step += 1
                                                                                           epoch: 15, tr loss: 0.004
                                                                                           epoch : 20, tr loss : 0.002
   else:
                                                                                           epoch: 25, tr loss: 0.001
       avg tr loss /= tr step
                                                                                           epoch: 30, tr loss: 0.001
       tr loss hist.append(avg tr loss)
   if (epoch + 1) \% 5 == 0:
       print('epoch : {:3}, tr_loss : {:.3f}'.format(epoch + 1, avg_tr_loss))
```

### Checking performance

```
yhat = model.predict(x_data)
yhat = np.argmax(yhat, axis=-1) * x_data_mask

pprint(list(map(lambda row : [idx2pos.get(elm) for elm in row],yhat.astype(np.int32).tolist())), width = 120)
pprint(pos)
```

```
[['pronoun', 'verb', 'adjective', '<pad>', '<pad>',
```



#### What's Next?

many to many bidirectional