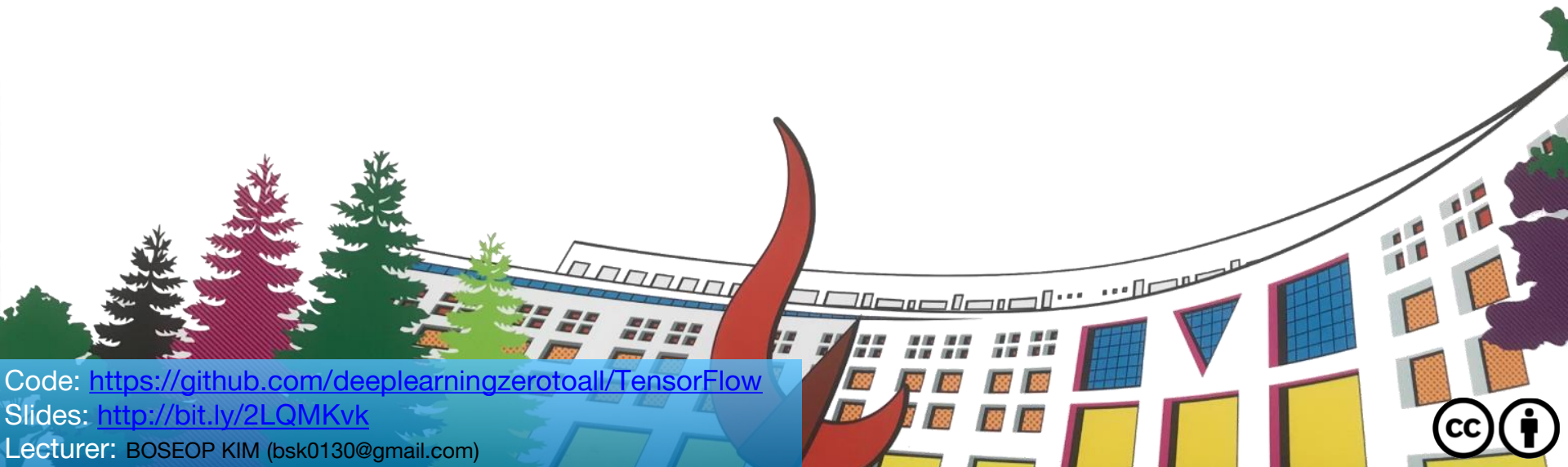


ML/DL for Everyone Season2

with  TensorFlow

Lab 12-1 many to one



Code: <https://github.com/deeplearningzerotoall/TensorFlow>

Slides: <http://bit.ly/2LQMKvk>

Lecturer: BOSEOP KIM (bsk0130@gmail.com)

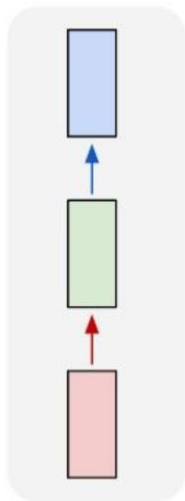


many to one

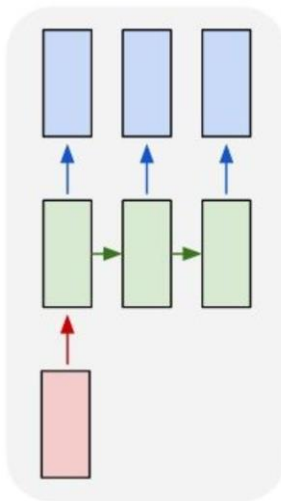
- Various usage of RNN
- What is “many to one”?
- Example : word sentiment classification
 - Preparing dataset
 - Creating and training model
 - Checking performance

Various usage of RNN

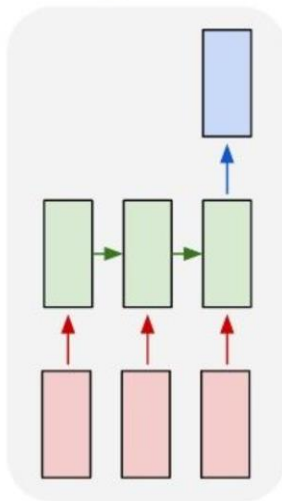
one to one



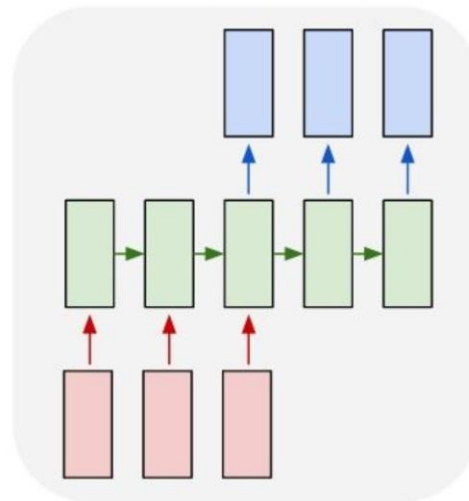
one to many



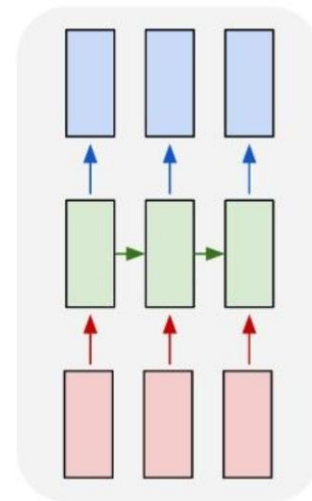
many to one



many to many



many to many



What is “many to one”?

Sequence classification

eg. classify polarity of sentence

sequence : sentence, tokens : word

['This movie is good']

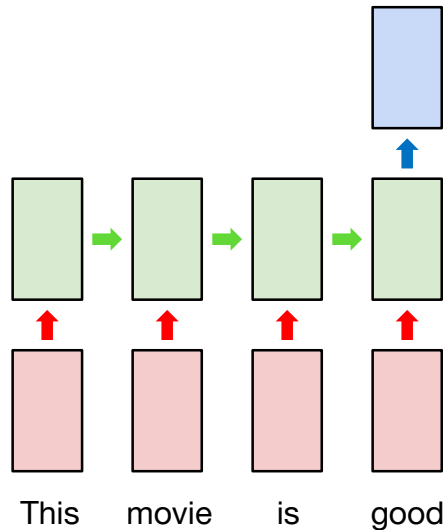
↓ *Tokenization*

['This', 'movie', 'is', 'good']

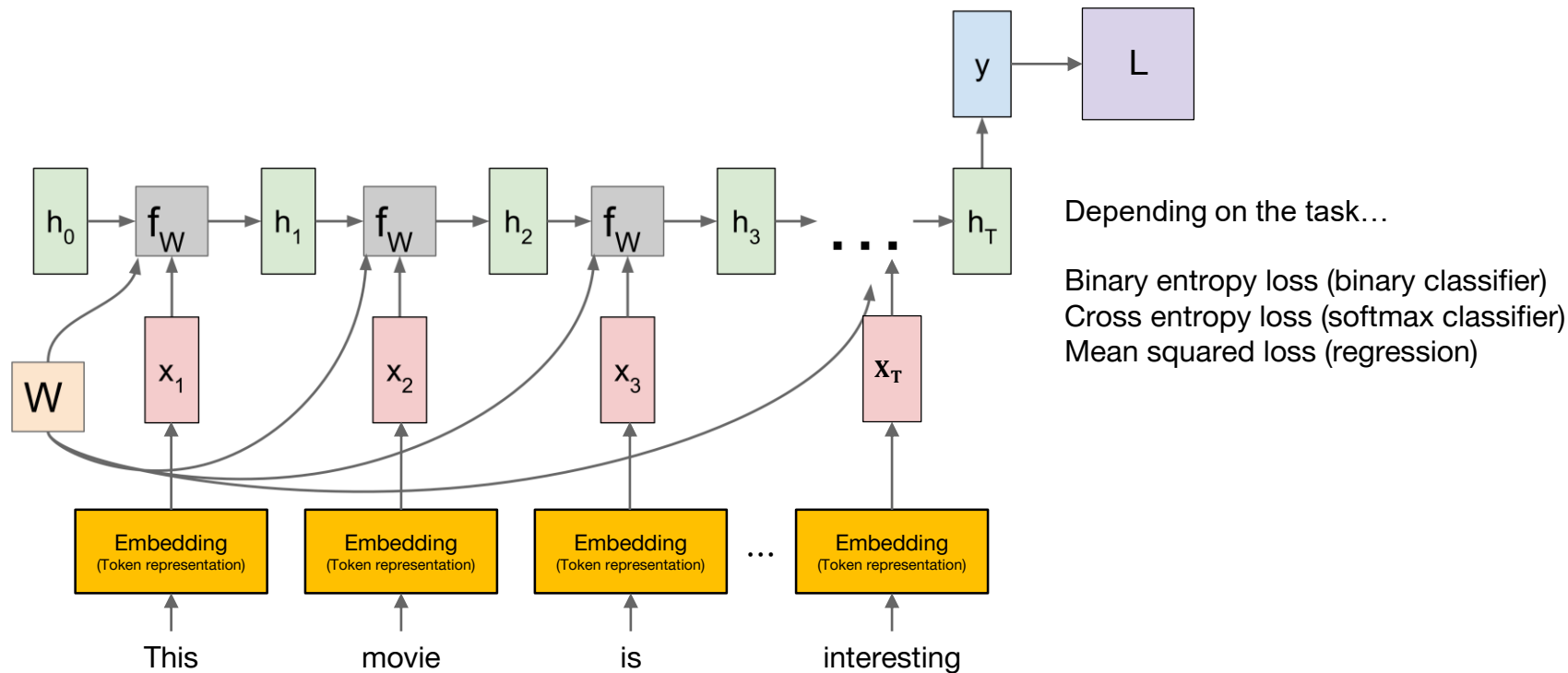
↓ *Classification*

Positive

Classification : **Positive** or **negative**?



What is “many to one”?



Example : word sentiment classification

Preparing dataset

example data

```
words = ['good', 'bad', 'worse', 'so good']
```

```
y_data = [1,0,0,1]
```

creating a token dictionary

```
char_set = ['<pad>'] + sorted(list(set(''.join(words))))
```

```
idx2char = {idx : char for idx, char in enumerate(char_set)}
```

```
char2idx = {char : idx for idx, char in enumerate(char_set)}
```

```
print(char_set)
```

```
print(idx2char)
```

```
print(char2idx)
```

```
['<pad>', ' ', 'a', 'b', 'd', 'e', 'g', 'o', 'r', 's', 'w']
```

```
{0: '<pad>', 1: ' ', 2: 'a', 3: 'b', 4: 'd', 5: 'e', 6: 'g', 7: 'o', 8: 'r', 9: 's', 10: 'w'}
```

```
{'<pad>': 0, ' ': 1, 'a': 2, 'b': 3, 'd': 4, 'e': 5, 'g': 6, 'o': 7, 'r': 8, 's': 9, 'w': 10}
```

Example : word sentiment classification

Preparing dataset

```
# converting sequence of tokens to sequence of indices
```

```
x_data = list(map(lambda word : [char2idx.get(char) for char in word], words))
```

```
x_data_len = list(map(lambda word : len(word), x_data))
```

```
print(x_data)
```

```
print(x_data_len)
```

```
[[6, 7, 7, 4], [3, 2, 4], [10, 7, 8, 9, 5], [9, 7, 1, 6, 7, 7, 4]]  
[4, 3, 5, 7]
```

```
# padding the sequence of indices
```

```
max_sequence = 10
```

```
x_data = pad_sequences(sequences = x_data, maxlen = max_sequence,  
                        padding = 'post', truncating = 'post')
```

```
# checking data
```

```
print(x_data)
```

```
print(x_data_len)
```

```
print(y_data)
```

```
[[ 6  7  7  4  0  0  0  0  0  0]  
[ 3  2  4  0  0  0  0  0  0  0]  
[10  7  8  9  5  0  0  0  0  0]  
[ 9  7  1  6  7  7  4  0  0  0]]  
[4, 3, 5, 7]  
[1, 0, 0, 1]
```

Example : word sentiment classification

Creating and training model

```
# creating simple rnn for "many to one" classification
```

```
input_dim = len(char2idx)
output_dim = len(char2idx)
one_hot = np.eye(len(char2idx))
hidden_size = 10
num_classes = 2
```

```
model = Sequential()
model.add(layers.Embedding(input_dim=input_dim, output_dim=output_dim,
                           trainable=False, mask_zero=True, input_length=max_sequence,
                           embeddings_initializer=keras.initializers.Constant(one_hot)))
model.add(layers.SimpleRNN(units=hidden_size))
model.add(layers.Dense(units=num_classes))
model.summary()
```

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 10, 11)	121
simple_rnn (SimpleRNN)	(None, 10)	220
dense (Dense)	(None, 2)	22
Total params: 363		
Trainable params: 242		
Non-trainable params: 121		

Example : word sentiment classification

Creating and training model

```
# creating loss function
def loss_fn(model, x, y):
    return tf.losses.sparse_softmax_cross_entropy(labels=y, logits=model(x))

# creating an optimizer
lr = .01
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))
tr_dataset = tr_dataset.shuffle(buffer_size = 4)
tr_dataset = tr_dataset.batch(batch_size = batch_size)

print(tr_dataset)

<BatchDataset shapes: ((?, 10), (?,)), types: (tf.int32, tf.int32)>
```

Example : word sentiment classification

Creating and training model

```
# training
tr_loss_hist = []
```

```
for epoch in range(epochs):
    avg_tr_loss = 0
    tr_step = 0
```

```
    for x_mb, y_mb in tr_dataset:
        with tf.GradientTape() as tape:
            tr_loss = loss_fn(model, x=x_mb, y=y_mb)
            grads = tape.gradient(target=tr_loss, sources=model.variables)
            opt.apply_gradients(grads_and_vars=zip(grads, model.variables))
            avg_tr_loss += tr_loss
            tr_step += 1
    else:
        avg_tr_loss /= tr_step
        tr_loss_hist.append(avg_tr_loss)
```

```
if (epoch + 1) % 5 == 0:
    print('epoch : {:3}, tr_loss : {:.3f}'.format(epoch + 1, avg_tr_loss))
```

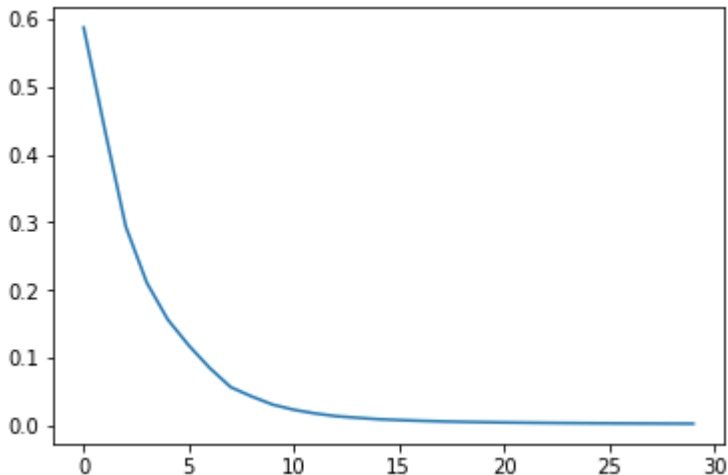
```
epoch :    5, tr_loss : 0.156
epoch :   10, tr_loss : 0.031
epoch :   15, tr_loss : 0.009
epoch :   20, tr_loss : 0.005
epoch :   25, tr_loss : 0.003
epoch :   30, tr_loss : 0.003
```

Example : word sentiment classification

Checking performance

```
yhat = model.predict(x_data)
yhat = np.argmax(yhat, axis=-1)
print('acc : {:.2%}'.format(np.mean(yhat == y_data)))
```

accuracy : 100.00%



What's Next?

- many to one stacking