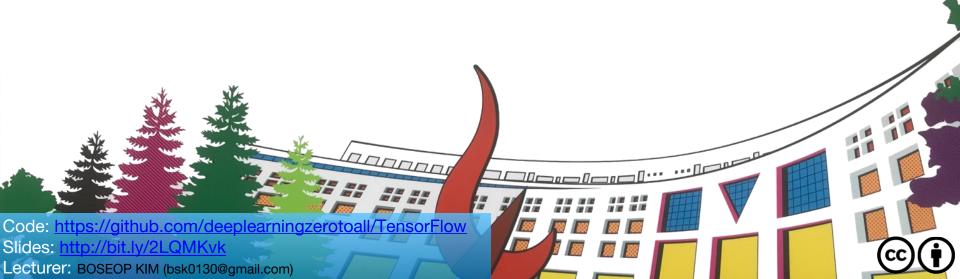
ML/DL for Everyone Season2



Lab 12-2 many to one stacking

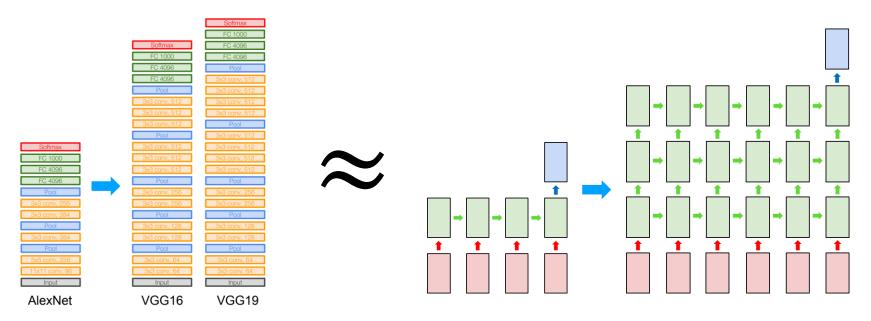


many to one stacking

- What is "stacking"?
- many to one stacking
- Example : sentence classification
 - Preparing dataset
 - Creating and training model
 - Checking performance

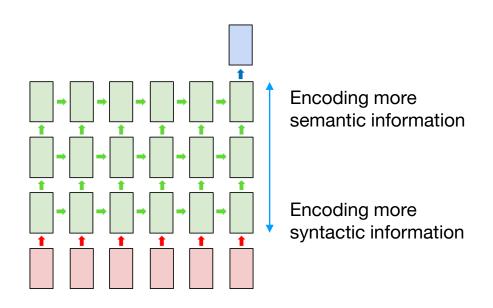
What is "stacking"?

While it is not theoretically clear what is the additional power gained by the deeper architecture, it was observed empirically that deep RNNs work better than shallower ones on some tasks.

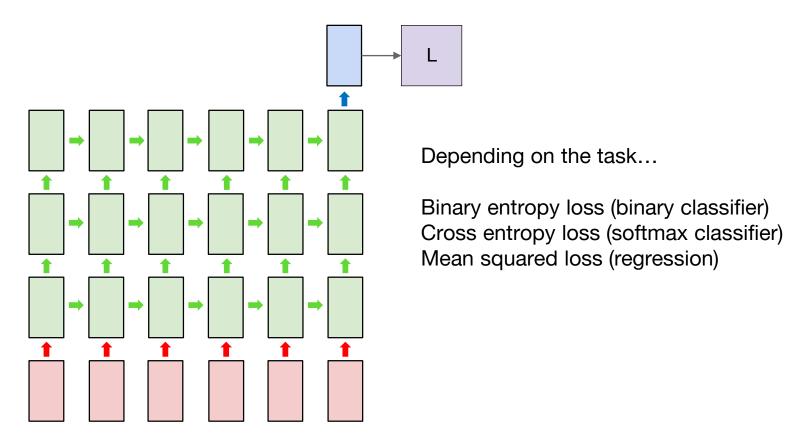


What is "stacking"?

Besides, many works have shown that different layers of deep RNNs encode different types of information.



many to one stacking



Preparing dataset

```
# example data
sentences = ['What I cannot create, I do not understand.',
            'Intellecuals solve problems, geniuses prevent them',
            'A person who never made a mistake never tied anything new.',
            'The same equations have the same solutions.']
v data = [1,0,0,1] # 1: richard feynman, 0: albert einstein
                                                                    ['<pad>', ' ', ',', '.', 'A', 'I', 'T', 'W', 'a',
# creating a token dictionary
                                                                    'b'. 'c'. 'd'. 'e', 'g', 'h', 'i', 'k', 'l', 'm',
char set = ['<pad>'] + sorted(list(set(''.join(sentences))))
                                                                    'n', 'o', 'p', 'q', 'r', 's', 't', 'u', 'v', 'w',
idx2char = {idx : char for idx, char in enumerate(char set)}
char2idx = {char : idx for idx, char in enumerate(char set)}
                                                                   {0: '<pad>', 1: ' ', 2: ',', 3: '.', 4: 'A', 5: 'I',
                                                                   6: 'T', 7: 'W', 8: 'a', 9: 'b', 10: 'c', 11: 'd', 12:
print(char set)
                                                                    'e', 13: 'g', 14: 'h', 15: 'i', 16: 'k', 17: 'l', 18:
print(idx2char)
                                                                    'm', 19: 'n', 20: 'o', 21: 'p', 22: 'q', 23: 'r', 24:
print(char2idx)
                                                                    's', 25: 't', 26: 'u', 27: 'v', 28: 'w', 29: 'v'}
                                                                   {'<pad>': 0, ' ': 1, ',': 2, '.': 3, 'A': 4, 'I': 5,
                                                                    'T': 6, 'W': 7, 'a': 8, 'b': 9, 'c': 10, 'd': 11,
                                                                    'e': 12, 'g': 13, 'h': 14, 'i': 15, 'k': 16, 'l': 17,
                                                                    'm': 18, 'n': 19, 'o': 20, 'p': 21, 'q': 22, 'r': 23,
                                                                    's': 24, 't': 25, 'u': 26, 'v': 27, 'w': 28, 'v': 29}
```

Example : sentence classificationPreparing dataset

```
# converting sequence of tokens to sequence of indices
x_data = list(map(lambda sentence : [char2idx.get(char) for char in sentence], sentences))
x_data_len = list(map(lambda sentence : len(sentence), sentences))

print(x_data)
print(x_data_len)
print(y_data)

[[7, 14, 8, 25, 1, 5, 1, 10, 8, 19, 19, 20, 25, 1, 10, 23, 12, 8, 25, 12, 2, 1, 5, 1, 11, 20, 1, 19, 20, 25, 1, 26, 19, 11, 12, 23, 24, 25, 8, 19, 11, 3], [5, 19, 25, 12, 17, 17, 12, 10, 26, 8, 17, 24, 1, 24, 20, 17, 27, 12, 1, 21, 23, 20, 9, 17, 12, 18, 24, 2, 1, 13, 12, 19, 15, 26, 24, 12, 24, 1, 21, 23, 12, 27, 12, 19, 25, 1, 25, 14, 12, 18], [4, 1, 21, 12, 23, 24, 20, 19, 1, 28, 14, 20, 1, 19, 12, 27, 12, 23, 1, 18, 8, 11, 12, 1, 8, 1, 18, 15, 24, 25, 8, 16, 12, 1, 19, 12, 27, 23, 1, 25, 15, 12, 11, 1, 8, 19, 29, 25, 14, 15, 19, 13, 1, 19, 12, 28, 3], [6, 14, 12, 1, 24, 8, 18, 12, 1, 12, 22, 26, 8, 25, 15, 20, 19, 24, 1, 14, 8, 27, 12, 1, 25, 14, 12, 1, 24, 8, 18, 12, 1, 24, 20, 17, 26, 25, 15, 20, 19, 24, 3]]
[1, 0, 0, 1]
```

Example : sentence classificationPreparing dataset

```
# padding the sequence of indices
max sequence = 55
x data = pad sequences(sequences = x data, maxlen = max sequence,
                     padding = 'post', truncating = 'post')
# checking data
print(x data)
                                                                  1 10 8 19 19 20 25 1 10 23 12
print(x data len)
                                                      1 19 20 25 1 26 19 11 12 23 24 25 8 19 11
print(y data)
                                               [ 5 19 25 12 17 17 12 10 26 8 17 24 1 24 20 17 27 12 1 21 23 20
                                                12 18 24 2 1 13 12 19 15 26 24 12 24 1 21 23 12 27 12 19 25
                                               [ 4 1 21 12 23 24 20 19 1 28 14 20 1 19 12 27 12 23 1 18
                                                 8 1 18 15 24 25 8 16 12 1 19 12 27 12 23 1 25 15 12 11
                                                                                                 25 14 15 19 13 1 19]
                                                               8 18 12 1 12 22 26 8 25 15 20 19 24 1 14 8 27
                                                25 14 12 1 24 8 18 12 1 24 20 17 26 25 15 20 19 24 3
                                                                                                     [42, 50, 58, 43]
                                                                                                          [1, 0, 0, 1]
```

Creating and training model

```
# creating stacked rnn for "many to one" classification with dropout
num classes = 2
hidden dims = [10,10]
input dim = len(char2idx)
output dim = len(char2idx)
one hot = np.eye(len(char2idx))
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 dims[0], return sequences=True))
model.add(layers.TimeDistributed(layers.Dropout(rate = .2)))
model.add(layers.SimpleRNN(units=hidden dims[1]))
model.add(layers.Dropout(rate = .2))
model.add(layers.Dense(units=num classes))
model.summary()
```

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 55, 30)	900
simple_rnn (SimpleRNN)	(None, 55, 10)	410
time_distributed (TimeDistri	(None, 55, 10)	0
simple_rnn_1 (SimpleRNN)	(None, 10)	210
dropout_1 (Dropout)	(None, 10)	0
dense (Dense)	(None, 2)	22
Total parama: 1 E42		

Total params: 1,542 Trainable params: 642 Non-trainable params: 900

Creating and training model

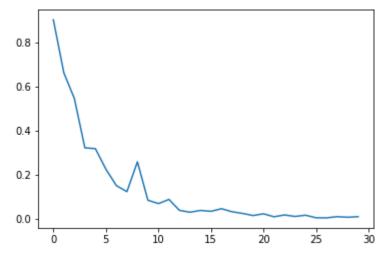
```
# creating loss function
def loss fn(model, x, y, training):
   return tf.losses.sparse softmax cross entropy(labels=y, logits=model(x, training))
# creating and optimizer
1r = .01
epochs = 30
batch size = 2
opt = tf.train.AdamOptimizer(learning rate = lr)
# generating data pipeline
tr dataset = 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: ((?, 55), (?,)), types: (tf.int32, tf.int32)>
```

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 in tr dataset:
       with tf.GradientTape() as tape:
           tr loss = loss fn(model, x=x mb, y=y mb, training=True)
       grads = tape.gradient(target=tr loss, sources=model.variables)
                                                                                              5, tr loss: 0.319
       opt.apply gradients(grads and vars=zip(grads, model.variables))
                                                                                     epoch: 10, tr loss: 0.084
       avg tr loss += tr loss
                                                                                     epoch: 15, tr loss: 0.038
       tr step += 1
                                                                                     epoch : 20, tr loss : 0.015
                                                                                     epoch : 25, tr_loss : 0.016
   else:
                                                                                     epoch : 30, tr loss : 0.010
       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))
```

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 many