In [1]:

```
from keras.datasets import imdb
from keras.preprocessing import sequence

max_features = 10000
max_len = 500

print('데이터 로드...')
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
print(len(x_train), '훈련 시퀀스')
print(len(x_train), '태스트 시퀀스')

print('시퀀스 패딩 (sample x time)')
x_train = sequence.pad_sequences(x_train, maxlen=max_len)
x_test = sequence.pad_sequences(x_test, maxlen=max_len)
print('x_train 크기:', x_train.shape)
print('x_test 크기:', x_test.shape)
```

Using TensorFlow backend.

```
데이터 로드...
25000 훈련 시퀀스
25000 테스트 시퀀스
시퀀스 패딩 (sample x time)
x_train 크기: (25000, 500)
x_test 크기: (25000, 500)
```

In [2]:

```
from keras.models import Sequential
from keras import layers
from keras.optimizers import RMSprop

model = Sequential()
model.add(layers.Embedding(max_features, 128, input_length=max_len))
model.add(layers.Conv1D(32, 7, activation='relu'))
model.add(layers.MaxPooling1D(5))
model.add(layers.Conv1D(32, 7, activation='relu'))
model.add(layers.GlobalMaxPooling1D())
model.add(layers.Dense(1))

model.summary()

model.compile(optimizer=RMSprop(Ir=1e-4), loss='binary_crossentropy', metrics=['acc'])
history = model.fit(x_train, y_train, epochs=10, batch_size=128, validation_split = 0.2)
```

Instructions for updating:

Colocations handled automatically by placer.

Layer (type)	Output	Shape	Param #
embedding_1 (Embedding)	(None,	500, 128)	1280000
conv1d_1 (Conv1D)	(None,	494, 32)	28704
max_pooling1d_1 (MaxPooling1	(None,	98, 32)	0
conv1d_2 (Conv1D)	(None,	92, 32)	7200
global_max_pooling1d_1 (Glob	(None,	32)	0
dense_1 (Dense)	(None,	1)	33

Total params: 1,315,937 Trainable params: 1,315,937 Non-trainable params: 0

WARNING:tensorflow:From C:\(\text{WProgramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packagesWtensorflow\(\text{WprogramDataWAnaconda3WIibWsite-packages\(\text{Wtensorflow}\); but on \(\text{WprogramDataWAnaconda3WIibWsite-packages\(\text{Wtensorflow}\); but on \(\text

Instructions for updating:

Use tf.cast instead.

WARNING:tensorflow:From C:\ProgramData\Anaconda3\Iib\site-packages\tensorflow\python\tensorflow.python.ops.math_ops) is deprecated a nd will be removed in a future version.

```
Instructions for updating:
Deprecated in favor of operator or tf.math.divide.
Train on 20000 samples, validate on 5000 samples
20000/20000 [=======] - 7s 355us/step - loss: 0.8337 - acc:
0.5093 - val_loss: 0.6874 - val_acc: 0.5646
Fnoch 2/10
0.6384 - val_loss: 0.6641 - val_acc: 0.6574
Epoch 3/10
20000/20000 [===========] - 1s 72us/step - loss: 0.6235 - acc:
0.7524 - val loss: 0.6078 - val acc: 0.7442
Epoch 4/10
20000/20000 [===========] - 1s 73us/step - loss: 0.5257 - acc:
0.8079 - val_loss: 0.4843 - val_acc: 0.8058
Epoch 5/10
20000/20000 [=========] - 1s 73us/step - loss: 0.4112 - acc:
0.8482 - val loss: 0.4276 - val acc: 0.8306
Epoch 6/10
20000/20000 [============ ] - 1s 73us/step - loss: 0.3462 - acc:
0.8672 - val_loss: 0.4146 - val_acc: 0.8394
Epoch 7/10
20000/20000 [=======] - 1s 73us/step - loss: 0.3071 - acc:
0.8669 - val_loss: 0.4365 - val_acc: 0.8250
Epoch 8/10
0.8531 - val_loss: 0.4272 - val_acc: 0.8054
Epoch 9/10
```

20000/20000 [========] - 1s 74us/step - loss: 0.2539 - acc:

```
0.8323 - val_loss: 0.4421 - val_acc: 0.7876
```

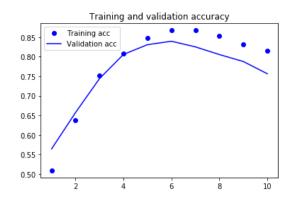
Epoch 10/10

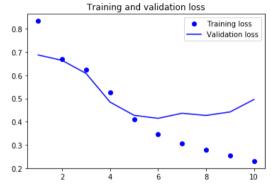
20000/20000 [======] - 1s 74us/step - loss: 0.2305 - acc:

0.8157 - val_loss: 0.4959 - val_acc: 0.7564

In [4]:

```
import matplotlib.pyplot as plt
acc = history.history['acc']
val_acc = history.history['val_acc']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(1, len(acc) + 1)
plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('Training and validation accuracy')
plt.legend()
plt.figure()
plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()
plt.show()
```





In [9]:

```
import os

data_dir = './jena_climate'
fname = os.path.join(data_dir, 'jena_climate_2009_2016.csv')

f = open(fname)
data = f.read()
f.close()
lines = data.split('\m')
header = lines[0].split(',')
lines = lines[1:]

print(header)
print(len(lines))
```

```
['"Date Time"', '"p (mbar)"', '"T (degC)"', '"Tpot (K)"', '"Tdew (degC)"', '"rh (%)"', '"VPmax (mbar)"', '"VPact (mbar)"', '"VPdef (mbar)"', '"sh (g/kg)"', '"H20C (mmol/mol)"', '"rho (g/m**3)"', '"wv (m/s)"', '"max. wv (m/s)"', '"wd (deg)"'] 420551
```

In [10]:

```
import numpy as np
float_data = np.zeros((len(lines), len(header) - 1))
for i, line in enumerate(lines):
    values = [float(x) for x in line.split(',')[1:]]
    float_data[i, :] = values
```

In [11]:

```
mean = float_data[:200000].mean(axis=0)
float_data -= mean
std = float_data[:200000].std(axis=0)
float_data /= std
```

In [12]:

```
def generator(data, lookback, delay, min_index, max_index, shuffle=False, batch_size=128, step=6
   if max index is None:
       max index = len(data) - delay - 1
   i = min index + lookback
   while 1:
        if shuffle:
           rows = np.random.randint(min_index + lookback, max_index, size=batch_size)
           if i + batch_size >= max_index:
               i = min index + lookback
           rows = np.arange(i. min(i + batch size. max index))
           i += len(rows)
        samples = np.zeros((len(rows), lookback // step, data.shape[-1]))
        targets = np.zeros((len(rows).))
        for j, row in enumerate(rows):
           indices = range(rows[i] - lookback, rows[i], step)
           samples[i] = data[indices]
           targets[i] = data[rows[i] + delav][1]
       yield samples, targets
```

In [14]:

1D 컨브넷 훈련 평가

In [15]:

```
from keras.models import Sequential
from keras import layers
from keras.optimizers import RMSprop

model = Sequential()
model.add(layers.Conv1D(32, 5, activation='relu', input_shape=(None, float_data.shape[-1])))
model.add(layers.MaxPooling1D(3))
model.add(layers.Conv1D(32, 5, activation='relu'))
model.add(layers.MaxPooling1D(3))
model.add(layers.Conv1D(32, 5, activation='relu'))
model.add(layers.GlobalMaxPooling1D())
model.add(layers.Dense(1))

model.summary()

model.compile(optimizer=RMSprop(), loss='mae')
history = model.fit_generator(train_gen, steps_per_epoch = 500, epochs=20, validation_data = val_gen, validation_steps = val_steps)
```

Layer (type)	Output	Shape		Param #	_
conv1d_3 (Conv1D)	(None,	None,	32)	2272	=
max_pooling1d_2 (MaxPooling1	(None,	None,	32)	0	-
conv1d_4 (Conv1D)	(None,	None,	32)	5152	_
max_pooling1d_3 (MaxPooling1	(None,	None,	32)	0	_
conv1d_5 (Conv1D)	(None,	None,	32)	5152	_
global_max_pooling1d_2 (Glob	(None,	32)		0	_
dense_2 (Dense)	(None,	1)		33	=
Total params: 12,609 Trainable params: 12,609 Non-trainable params: 0					_
Epoch 1/20 500/500 [===================================		====]	- 14s	28ms/step - loss:	0.4197 - val_los
500/500 [====== s: 0.4546		====]	- 14s	28ms/step - loss:	0.3647 - val_los
Epoch 3/20 500/500 [====== s: 0.4752	=====	====]	- 14s	28ms/step - loss:	0.3414 - val_los
Epoch 4/20 500/500 [======= s: 0.4701		====]	- 14s	28ms/step - loss:	0.3266 - val_los
Epoch 5/20 500/500 [======= s: 0.4686		====]	- 14s	28ms/step - loss:	0.3173 - val_los
Epoch 6/20 500/500 [======= s: 0.4930		====]	- 14s	27ms/step - loss:	0.3048 - val_los
Epoch 7/20 500/500 [======= s: 0.4961		====]	- 14s	28ms/step - loss:	0.2971 - val_los
Epoch 8/20 500/500 [======= s: 0.5057		====]	- 14s	28ms/step - loss:	0.2898 - val_los
Epoch 9/20 500/500 [====== s: 0.4800		====]	- 14s	28ms/step - loss:	0.2831 - val_los
Epoch 10/20 500/500 [======= s: 0.4844		====]	- 14s	28ms/step - loss:	0.2787 - val_los
Epoch 11/20 500/500 [======= s: 0.5068		====]	- 14s	28ms/step - loss:	0.2749 - val_los
Epoch 12/20 500/500 [======= s: 0.4961		====]	- 14s	28ms/step - loss:	0.2695 - val_los
Epoch 13/20 500/500 [======= s: 0.5277 Epoch 14/20		====]	- 14s	28ms/step - loss:	0.2643 - val_los

```
500/500 [========] - 14s 28ms/step - loss: 0.2619 - val_los
s: 0.5170
Epoch 15/20
500/500 [=======] - 14s 28ms/step - loss: 0.2603 - val los
s: 0.4916
Epoch 16/20
500/500 [======] - 14s 28ms/step - loss: 0.2558 - val los
s: 0.5125
Epoch 17/20
500/500 [========] - 14s 28ms/step - loss: 0.2540 - val_los
s: 0.5049
Epoch 18/20
500/500 [===
         s: 0.4988
Fpoch 19/20
500/500 [===
        s: 0.5116
Epoch 20/20
500/500 [======] - 14s 28ms/step - loss: 0.2483 - val_los
s: 0.5070
```

In [16]:

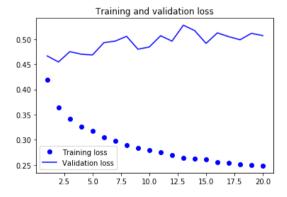
```
import matplotlib.pyplot as plt

loss = history.history['loss']
val_loss = history.history['val_loss']

epochs = range(1, len(loss) + 1)

plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()

plt.show()
```



합성곱과 GRU 연결

In [17]:

```
from keras.layers import GRU

model = Sequential()
model.add(layers.Conv1D(32, 5, activation='relu', input_shape=(None, float_data.shape[-1])))
model.add(layers.MaxPooling1D(3))
model.add(layers.Conv1D(32, 5, activation='relu'))
model.add(layers.GRU(32, dropout=0.1, recurrent_dropout=0.5))
model.add(layers.Dense(1))

model.summary()
model.compile(optimizer=RMSprop(), loss='mae')
history = model.fit_generator(train_gen, steps_per_epoch = 500, epochs=20, validation_data = val_gen, validation_steps = val_steps)
```

Instructions for updating:

Please use 'rate' instead of 'keep_prob'. Rate should be set to 'rate = $1 - \text{keep_p}$ rob'.

Layer (type)	Output	Shape	Param #
conv1d_6 (Conv1D)	(None,	None, 32)	2272
max_pooling1d_4 (MaxPooling1	(None,	None, 32)	0
conv1d_7 (Conv1D)	(None,	None, 32)	5152
gru_1 (GRU)	(None,	32)	6240
dense_3 (Dense)	(None,	1)	33

Total params: 13,697 Trainable params: 13,697 Non-trainable params: 0

Fpoch 1/20 500/500 [== ======] - 67s 135ms/step - loss: 0.3479 - val_los s: 0.3040 Epoch 2/20 500/500 [== s: 0.3261 Epoch 3/20 500/500 [=== s: 0.2889 Epoch 4/20 500/500 [=== s: 0.3042 Epoch 5/20 500/500 [======] - 67s 134ms/step - loss: 0.2806 - val los s: 0.2879 Epoch 6/20 s: 0.2855 Epoch 7/20 500/500 [=== s: 0.3166 Epoch 8/20 s: 0.2948 Epoch 9/20 s: 0.2989 Epoch 10/20 500/500 [=== s: 0.2917 Epoch 11/20 500/500 [==: s: 0.2960 Epoch 12/20 500/500 [==: =========] - 66s 132ms/step - loss: 0.2438 - val_los s: 0.3001 Epoch 13/20

```
s: 0.2974
Fnoch 14/20
s: 0.3019
Epoch 15/20
s: 0.3020
Epoch 16/20
500/500 [======] - 65s 130ms/step - loss: 0.2298 - val los
s: 0.3009
Epoch 17/20
s: 0.3088
Fnoch 18/20
500/500 [====
        s: 0.3074
Epoch 19/20
s: 0.3016
Fpoch 20/20
500/500 [===
          =========] - 65s 131ms/step - loss: 0.2208 - val_los
s: 0.3025
```

In [18]:

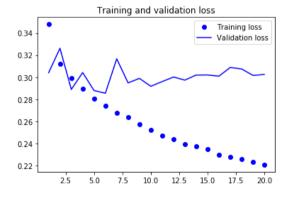
```
import matplotlib.pyplot as plt

loss = history.history['loss']
val_loss = history.history['val_loss']

epochs = range(1, len(loss) + 1)

plt.plot(epochs, loss, 'bo', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend()

plt.show()
```



In []: