

In [1]:

```
import numpy as np

from tensorflow.keras.utils import to_categorical
from tensorflow.keras.datasets import imdb
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import SimpleRNN, Dense
from tensorflow.keras.optimizers import Adam

import matplotlib.pyplot as plt
```

In [2]:

```
(x_train, t_train), (x_test, t_test) = imdb.load_data(num_words = 1000)

print("x_train shape : ", x_train.shape)
print("t_train shape : ", t_train.shape)

print("x_test shape : ", x_test.shape)
print("t_test shape : ", t_test.shape)
```

```
x_train shape : (25000,)
t_train shape : (25000,)
x_test shape : (25000,)
t_test shape : (25000,)
```

In [3]:

```
x_train = x_train[:5000]
t_train = t_train[:5000]

x_test = x_test[:1000]
t_test = t_test[:1000]

print("개수 변경 후 x_train shape : ", x_train.shape)
print("개수 변경 후 t_train shape : ", t_train.shape)

print("개수 변경 후 x_test shape : ", x_test.shape)
print("개수 변경 후 t_test shape : ", t_test.shape)

num_classes= len(set(t_train))
print("정답 레이블 종류 : {}".format(num_classes))
```

```
개수 변경 후 x_train shape : (5000,)
개수 변경 후 t_train shape : (5000,)
개수 변경 후 x_test shape : (1000,)
개수 변경 후 t_test shape : (1000,)
정답 레이블 종류 : 2
```

In [4]:

```
print("train용 1번째 리뷰 / 판별 : ", x_train[0], "/", t_train[0])
print(" ")
print("test용 1번째 리뷰 / 판별 : ", x_test[0], "/", t_test[0])
```

train용 1번째 리뷰 / 판별 : [1, 14, 22, 16, 43, 530, 973, 2, 2, 65, 458, 2, 66, 2, 4, 173, 36, 256, 5, 25, 100, 43, 838, 112, 50, 670, 2, 9, 35, 480, 284, 5, 150, 4, 172, 112, 167, 2, 336, 385, 39, 4, 172, 2, 2, 17, 546, 38, 13, 447, 4, 192, 50, 16, 6, 147, 2, 19, 14, 22, 4, 2, 2, 469, 4, 22, 71, 87, 12, 16, 43, 530, 38, 76, 15, 13, 2, 4, 22, 17, 515, 17, 12, 16, 626, 18, 2, 5, 62, 386, 12, 8, 316, 8, 106, 5, 4, 2, 2, 16, 480, 66, 2, 33, 4, 130, 12, 16, 38, 619, 5, 25, 124, 51, 36, 135, 48, 25, 2, 33, 6, 22, 12, 215, 28, 77, 52, 5, 14, 407, 16, 82, 2, 8, 4, 107, 117, 2, 15, 256, 4, 2, 7, 2, 5, 723, 36, 71, 43, 530, 476, 26, 400, 317, 46, 7, 4, 2, 2, 13, 104, 88, 4, 381, 15, 297, 98, 32, 2, 56, 26, 141, 6, 194, 2, 18, 4, 226, 22, 21, 134, 476, 26, 480, 5, 144, 30, 2, 18, 51, 36, 28, 224, 92, 25, 104, 4, 226, 65, 16, 38, 2, 88, 12, 16, 283, 5, 16, 2, 113, 103, 32, 15, 16, 2, 19, 178, 32] / 1

test용 1번째 리뷰 / 판별 : [1, 591, 202, 14, 31, 6, 717, 10, 10, 2, 2, 5, 4, 360, 7, 4, 177, 2, 394, 354, 4, 123, 9, 2, 2, 2, 10, 10, 13, 92, 124, 89, 488, 2, 100, 2, 8, 2, 14, 31, 23, 27, 2, 29, 220, 468, 8, 124, 14, 286, 170, 8, 157, 46, 5, 27, 239, 16, 179, 2, 38, 32, 25, 2, 451, 202, 14, 6, 717] / 0

In [5]:

```
word_to_index = imdb.get_word_index()
print(word_to_index)
```

{ 'fawn': 34701, 'tsukino': 52006, 'nunnery': 52007, 'sonja': 16816, 'vani': 63951, 'woods': 1408, 'spiders': 16115, 'hanging': 2345, 'woody': 2289, 'trawling': 52008, "hold's": 52009, 'comically': 11307, 'localized': 40830, 'disobeying': 30568, "royale": 52010, "harpo's": 40831, 'canet': 52011, 'aileen': 19313, 'acurately': 52012, 'diplomat's': 52013, 'rickman': 25242, 'arranged': 6746, 'rumbustious': 52014, 'familiarness': 52015, "spider'": 52016, 'hahahah': 68804, "wood'": 52017, 'transvestism': 40833, "hangin'": 34702, 'bringing': 2338, 'seamier': 40834, 'wooded': 34703, 'bavora': 52018, 'grueling': 16817, 'wooden': 1636, 'wednesday': 16818, "'prix": 52019, 'altagracia': 34704, 'circuitry': 52020, 'crotch': 11585, 'busybody': 57766, "tart'n'tangy": 52021, 'burgade': 14129, 'thrace': 52023, "tom's": 11038, 'snuggles': 52025, 'francesco': 29114, 'complainers': 52027, 'templarios': 52125, '272': 40835, '273': 52028, 'zaniacs': 52130, '275': 34706, 'consenting': 27631, 'snuggled': 40836, 'inanimate': 15492, 'uality': 52030, 'bronte': 11926, 'errors': 4010, 'dialogs': 3230, "yomada's": 52031, "madman's": 34707, 'dialoge': 30585, 'usenet': 52033, 'videodrome': 40837, "kid'": 26338, 'pawed': 52034, "'girlfriend'": 30569, "'pleasure": 52035, "'reloaded'": 52036, "kazakos'": 40839, 'rocque': 52037, 'mailings': 52038, 'brainwashed': 11927, 'mcanally': 16819, "tom'": 52039, 'kurupt': 25243, 'affiliated': 21905, 'babaganoosh': 52040, 'noe's': 40840, 'quart': 40841, 'kids': 359, 'upliftin g': 5034, 'controversy': 7093, 'kida': 21906, 'kidd': 23379, 'error': 52041, 'neuro

In [6]:

```
index_to_word = dict([(value, key) for (key, value) in word_to_index.items()])
print(index_to_word)
```

```
{34701: 'fawn', 52006: 'tsukino', 52007: 'nunnery', 16816: 'sonja', 63951: 'vani', 1408: 'woods', 16115: 'spiders', 2345: 'hanging', 2289: 'woody', 52008: 'trawling', 52009: 'hold's', 11307: 'comically', 40830: 'localized', 30568: 'disobeying', 52010: 'royale', 40831: 'harpo's', 52011: 'canet', 19313: 'aileen', 52012: 'acurately', 52013: 'diplomat's', 25242: 'rickman', 6746: 'arranged', 52014: 'rumbustious', 52015: 'familiarness', 52016: 'spider'", 68804: 'hahahah', 52017: 'wood"', 40833: 'transvestism', 34702: 'hangin"', 2338: 'bringing', 40834: 'seamier', 34703: 'wooded', 52018: 'bravora', 16817: 'grueling', 1636: 'wooden', 16818: 'wednesday', 52019: 'prix', 34704: 'altagracia', 52020: 'circuitry', 11585: 'crotch', 57766: 'busybody', 52021: 'tart'n'tangy', 14129: 'burgade', 52023: 'thrace', 11038: 'tom's', 52025: 'snuggles', 29114: 'francesco', 52027: 'complainers', 52125: 'templarios', 40835: '272', 52028: '273', 52130: 'zaniacs', 34706: '275', 27631: 'consenting', 40836: 'snuggled', 15492: 'inanimate', 52030: 'uality', 11926: 'bronte', 4010: 'errors', 3230: 'dialogs', 52031: 'yomada's', 34707: 'madman's', 30585: 'dialoge', 52033: 'usenet', 40837: 'videodrome', 26338: 'kid"', 52034: 'pawed', 30569: 'girlfriend"', 52035: 'pleasure', 52036: 'reloaded"', 40839: 'kazakos"', 52037: 'rocque', 52038: 'mailings', 11927: 'brainwashed', 16819: 'mcanally', 52039: 'tom"', 25243: 'krupt', 21905: 'affiliated', 52040: 'babaganoosh', 40840: 'noe's', 40841: 'quart', 359: 'kids', 5034: 'uplifting', 7093: 'controversy', 21906: 'kida', 23379: 'kidd', 52041: 'error"', 52042: 'ne...
16516: 'hatched', 66576: 'hatched', 66576: 'hatched', 66576: 'hatched', 40842: 'feet from...
```

In [7]:

```
word_to_index.items()
```

Out[7]:

```
dict_items([('fawn', 34701), ('tsukino', 52006), ('nunnery', 52007), ('sonja', 16816), ('vani', 63951), ('woods', 1408), ('spiders', 16115), ('hanging', 2345), ('woody', 2289), ('trawling', 52008), ('hold's', 52009), ('comically', 11307), ('localized', 40830), ('disobeying', 30568), ('royale', 52010), ('harpo's', 40831), ('canet', 52011), ('aileen', 19313), ('acurately', 52012), ('diplomat's', 52013), ('rickman', 25242), ('arranged', 6746), ('rumbustious', 52014), ('familiarness', 52015), ('spider"', 52016), ('hahahah', 68804), ('wood"', 52017), ('transvestism', 40833), ('hangin"', 34702), ('bringing', 2338), ('seamier', 40834), ('wooded', 34703), ('bravora', 52018), ('grueling', 16817), ('wooden', 1636), ('wednesday', 16818), ('prix', 52019), ('altagracia', 34704), ('circuitry', 52020), ('crotch', 11585), ('busybody', 57766), ('tart'n'tangy', 52021), ('burgade', 14129), ('thrace', 52023), ('tom's', 11038), ('snuggles', 52025), ('francesco', 29114), ('complainers', 52027), ('templarios', 52125), ('272', 40835), ('273', 52028), ('zaniacs', 52130), ('275', 34706), ('consenting', 27631), ('snuggled', 40836), ('inanimate', 15492), ('uality', 52030), ('bronte', 11926), ('errors', 4010), ('dialogs', 3230), ('yomada's', 52031), ('madman's', 34707), ('dialoge', 30585), ('usenet', 52033), ('videodrome', 40837), ('kid"', 26338), ('pawed', 52034), ('girlfriend"', 30569), ('pleasure', 52035), ('reloaded"', 52036), ('kazakos"', 40839), ('rocque', 52037), ('mailings', 52038), ('brainwas...
```

In [8]:

```
print("빈도수 최상위 (1등) 단어 : {}".format(index_to_word[1]))
print("빈도수 상위 3938등 단어 : {}".format(index_to_word[3938]))
print("빈도수 최하위 (꼴등", str(len(word_to_index)), ") 단어 : {}".format(index_to_word[len(word_t
```

```
빈도수 최상위 (1등) 단어 : the
빈도수 상위 3938등 단어 : suited
빈도수 최하위 (꼴등 88584 )) 단어 : 'l'
```

In [9]:

```
print(" ".join([index_to_word[index] for index in x_train[0]]))
```

the as you with out themselves powerful and and their becomes and had and of lot from anyone to have after out atmosphere never more room and it so heart shows to years of every never going and help moments or of every and and movie except her was several of enough more with is now and film as you of and and unfortunately of you than him that with out themselves her get for was and of you movie sometimes movie that with scary but and to story wonderful that in seeing in character to of and and with heart had and they of here that with her serious to have does when from why what have and they is you that isn't one will very to as itself with other and in of seen over and for anyone of and br and to whether from than out themselves history he name half some br of and and was two most of mean for 1 any an and she he should is thought and but of script you not while history he heart to real at and but when from one bit then have two of script their with her and most that with wasn't to with and acting watch an for with and film want an

In [10]:

```
index_to_word = {}

for key, value in word_to_index.items():
    index_to_word[value + 3] = key

print("빈도수 최상위 (1등) 단어 : {}".format(index_to_word[4]))
print("빈도수 상위 3938등 단어 : {}".format(index_to_word[3941]))
print("빈도수 최하위 (꼴등", str(len(word_to_index)), ") 단어 : {}".format(index_to_word[len(word_t
```

빈도수 최상위 (1등) 단어 : the  
빈도수 상위 3938등 단어 : suited  
빈도수 최하위 (꼴등 88584 )) 단어 : 'I'

In [11]:

```
index_to_word
```

Out[11]:

```
{34704: 'fawn',
 52009: 'tsukino',
 52010: 'nunnery',
 16819: 'sonja',
 63954: 'vani',
 1411: 'woods',
 16118: 'spiders',
 2348: 'hanging',
 2292: 'woody',
 52011: 'trawling',
 52012: "hold's",
 11310: 'comically',
 40833: 'localized',
 30571: 'disobeying',
 52013: "'royale",
 40834: "harpo's",
 52014: 'canet',
 19316: 'aileen'.
```

In [12]:

```
for index, token in enumerate(["?", "?", "?", "?"]):  
    print("index, token : ", index, token)  
    index_to_word[index] = token
```

```
index, token : 0 ?  
index, token : 1 ?  
index, token : 2 ?  
index, token : 3 ?
```

In [13]:

```
for i in range(0, 4):  
    print(str(i) + "번째 index_to_word : {}".format(index_to_word[i]))
```

```
0번째 index_to_word : ?  
1번째 index_to_word : ?  
2번째 index_to_word : ?  
3번째 index_to_word : ?
```

In [14]:

```
print(" ".join([index_to_word[index] for index in x_train[0]]))
```

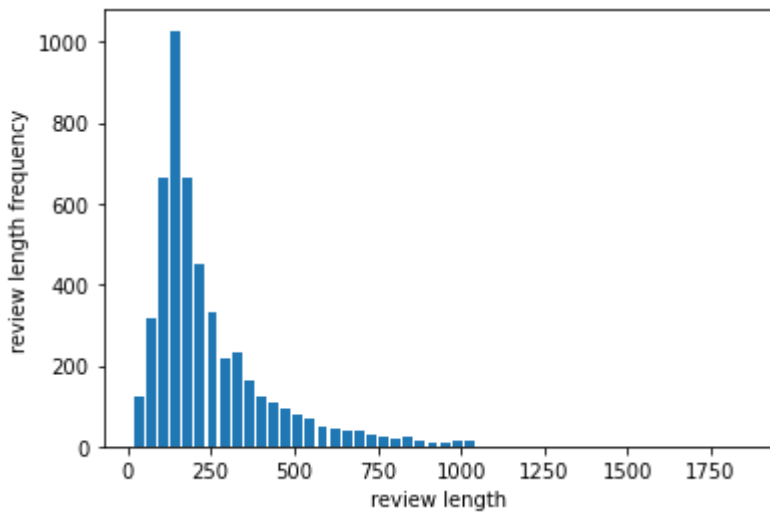
? this film was just brilliant casting ?? story direction ? really ? the part they played and you could just imagine being there robert ? is an amazing actor and now t he same being director ? father came from the same ?? as myself so i loved the fact there was a real ? with this film the ?? throughout the film were great it was just brilliant so much that i ? the film as soon as it was released for ? and would recom mend it to everyone to watch and the ?? was amazing really ? at the end it was so s ad and you know what they say if you ? at a film it must have been good and this def initely was also ? to the two little ? that played the ? of ? and paul they were jus t brilliant children are often left out of the ?? i think because the stars that pl ay them all ? up are such a big ? for the whole film but these children are amazing and should be ? for what they have done don't you think the whole story was so ? bec ause it was true and was ? life after all that was ? with us all

In [15]:

```
review_length = np.array([len(x) for x in x_train])
print("train용 리뷰 길이 : ", review_length)
print("리뷰의 최대 길이 : {}".format(np.max(review_length)))
print("리뷰의 최소 길이 : {}".format(np.min(review_length)))
print("리뷰의 평균 길이 : {}".format(np.mean(review_length)))

plt.hist(review_length, bins = 50, rwidth=0.8)
plt.xlabel("review length")
plt.ylabel("review length frequency")
plt.show()
```

```
train용 리뷰 길이 : [218 189 141 ... 852 130 171]
리뷰의 최대 길이 : 1851
리뷰의 최소 길이 : 16
리뷰의 평균 길이 : 243.8212
```



In [16]:

```
print("x_train shape : ", x_train.shape)
print("x_test shape : ", x_test.shape)

train_seq = pad_sequences(x_train, maxlen = 100)
test_seq = pad_sequences(x_test, maxlen = 100)

print("train_seq shape : ", train_seq.shape)
print("test_seq shape : ", test_seq.shape)
```

```
x_train shape : (5000,)
x_test shape : (1000,)
train_seq shape : (5000, 100)
test_seq shape : (1000, 100)
```

In [17]:

```
x_train
```

Out[17]:





```

91, 242, 47, 8, 81, 19, 2, 2, 21, 12, 9, 6, 2, 2, 8, 148, 11, 4, 124, 319, 2, 2, 11,
4, 984, 21, 64, 929, 7, 848, 2, 17, 2, 11, 4, 501, 10, 10, 444, 13, 423, 4, 123, 2,
18, 4, 2, 13, 62, 28, 2, 4, 405, 343, 11, 2, 2, 2, 21, 13, 70, 391, 138, 2, 181, 6,
print("Join([index, to_word[index] for index in x_train[0]])")
53, 65, 2, 418, 348, 4, 875, 551, 2, 13, 40, 134, 84, 45, 327, 8, 67, 98, 46, 5, 44,
131this film was just brilliant casting 722, story direction 657, really 69? the part they
played and you could just imagine being there, robert 501) an amazing actor and now t
he same being director? father came from the same 2, 2 as myself 377, i loved the fact
there was a real 5? with this film the 77? 69 throughout the film were great it was just
brilliant so much that 31? the film as soon as it was released for 435, and would recom
mend it to everyone to watch and the 20 was amazing really 18? 583 the end it was so s
ad and you know what they say if you 2, a 2 film it must have been good and this def
initely was also 17? to the two little 2? that played the 448 23 and 45 pad, they were jus
t brilliant children are often left out of the ? ? i think because the stars that pl
ay them last 11 up are such a big 10 for the whole film but these children are amazing
and should be 2 for what they have done 683, you think the whole story was 901? bec
ause it was true and was life after all 39, 6, 248? with 69, 268, 410, 62, 115, 30,
617, 13, 301, 252, 2, 6, 2, 7, 2, 920, 924, 969, 2, 2, 56, 49, 481, 302, 5, 446, 11,
160, 2, 2, 2, 8, 297, 4, 668, 173, 17, 59, 575, 8, 30, 23, 4, 945, 33, 4, 58, 59, 27
6, 23, 6, 196, 613, 2, 72, 33, 4, 2, 11, 4, 365, 673, 5, 75, 605, 4, 2, 11, 6, 686,
193, 2, 59, 95, 435, 145, 8, 41, 123, 137, 13, 2, 11, 4, 2, 481, 302, 2, 28, 115, 22
train_seq = to_categorical(train_seq)
test_seq = to_categorical(test_seq)
dtype=object)
print("One-hot Vector 적용 후 train_seq.shape : ", train_seq.shape)
print("One-hot Vector 적용 후 test_seq.shape : ", test_seq.shape)

```

One-hot Vector 적용 후 train\_seq.shape : (5000, 100, 1000)

One-hot Vector 적용 후 test\_seq.shape : (1000, 100, 1000)

In [23]:

```
cell_size = 128
timesteps = 100
feature = 1000

model = Sequential(name="IMDB_RNN")

model.add(SimpleRNN(cell_size, kernel_initializer="he_uniform",
                    input_shape = (timesteps, feature), activation = "relu",
                    return_sequences=True))

model.add(SimpleRNN(cell_size, kernel_initializer="he_uniform", activation = "relu"))

model.add(Dense(1, activation = "sigmoid"))

model.compile(loss="binary_crossentropy",
              optimizer=Adam(learning_rate=0.001, clipnorm=1.), metrics=["accuracy"])

model.summary()
```

Model: "IMDB\_RNN"

| Layer (type)              | Output Shape     | Param # |
|---------------------------|------------------|---------|
| =====                     |                  |         |
| simple_rnn (SimpleRNN)    | (None, 100, 128) | 144512  |
| simple_rnn_1 (SimpleRNN)  | (None, 128)      | 32896   |
| dense (Dense)             | (None, 1)        | 129     |
| =====                     |                  |         |
| Total params: 177,537     |                  |         |
| Trainable params: 177,537 |                  |         |
| Non-trainable params: 0   |                  |         |
| =====                     |                  |         |

In [24]:

```
cell_size = 128
timesteps = 100
feature = 1000

model = Sequential(name="Weight_test")

model.add(SimpleRNN(cell_size, kernel_initializer="he_uniform",
                    input_shape = (timesteps, feature),
                    activation = "relu", return_sequences=True))

print("모델의 가중치 (He Uniform 초기화 적용) : ", model.get_weights(), sep="\n")
```

모델의 가중치 (He Uniform 초기화 적용) :

[illegible]

In [25]:

```
cell_size = 128
timesteps = 100
feature = 1000

model = Sequential(name="IMDB_RNN")

model.add(SimpleRNN(cell_size, kernel_initializer="he_uniform",
                    input_shape = (timesteps, feature), activation = "relu",
                    return_sequences=True))

model.add(SimpleRNN(cell_size, kernel_initializer="he_uniform", activation = "relu"))

model.add(Dense(1, activation = "sigmoid"))

model.compile(loss="binary_crossentropy",
              optimizer=Adam(learning_rate=0.001, clipnorm=1.), metrics=["accuracy"])

model.summary()
```

Model: "IMDB\_RNN"

| Layer (type)              | Output Shape     | Param # |
|---------------------------|------------------|---------|
| =====                     |                  |         |
| simple_rnn_3 (SimpleRNN)  | (None, 100, 128) | 144512  |
| simple_rnn_4 (SimpleRNN)  | (None, 128)      | 32896   |
| dense_1 (Dense)           | (None, 1)        | 129     |
| =====                     |                  |         |
| Total params: 177,537     |                  |         |
| Trainable params: 177,537 |                  |         |
| Non-trainable params: 0   |                  |         |
| =====                     |                  |         |

In [26]:

```
model.fit(train_seq, t_train, epochs=20, batch_size = 64)
```

Epoch 1/20  
79/79 [=====] - 22s 221ms/step - loss: 0.6850 - accuracy:  
0.5404

Epoch 2/20  
79/79 [=====] - 16s 204ms/step - loss: 0.5843 - accuracy:  
0.7074

Epoch 3/20  
79/79 [=====] - 16s 202ms/step - loss: 0.5460 - accuracy:  
0.7806

Epoch 4/20  
79/79 [=====] - 16s 200ms/step - loss: 0.6525 - accuracy:  
0.7624

Epoch 5/20  
79/79 [=====] - 17s 220ms/step - loss: 1.1107 - accuracy:  
0.8010

Epoch 6/20  
79/79 [=====] - 16s 207ms/step - loss: 0.5616 - accuracy:  
0.8010

Epoch 7/20  
79/79 [=====] - 17s 212ms/step - loss: 0.8111 - accuracy:  
0.8144

Epoch 8/20  
79/79 [=====] - 20s 250ms/step - loss: 5.3154 - accuracy:  
0.8238

Epoch 9/20  
79/79 [=====] - 22s 273ms/step - loss: 1.1088 - accuracy:  
0.8696

Epoch 10/20  
79/79 [=====] - 20s 246ms/step - loss: 1.0905 - accuracy:  
0.8968

Epoch 11/20  
79/79 [=====] - 24s 301ms/step - loss: 0.1971 - accuracy:  
0.9394

Epoch 12/20  
79/79 [=====] - 25s 318ms/step - loss: 0.1638 - accuracy:  
0.9574

Epoch 13/20  
79/79 [=====] - 20s 250ms/step - loss: 0.0920 - accuracy:  
0.9728

Epoch 14/20  
79/79 [=====] - 19s 233ms/step - loss: 0.0719 - accuracy:  
0.9784

Epoch 15/20  
79/79 [=====] - 18s 232ms/step - loss: 0.1614 - accuracy:  
0.9768

Epoch 16/20  
79/79 [=====] - 20s 258ms/step - loss: 0.1721 - accuracy:  
0.9870

Epoch 17/20  
79/79 [=====] - 21s 259ms/step - loss: 0.0374 - accuracy:  
0.9906

Epoch 18/20  
79/79 [=====] - 19s 237ms/step - loss: 0.6445 - accuracy:  
0.9812

Epoch 19/20  
79/79 [=====] - 18s 224ms/step - loss: 0.0214 - accuracy:  
0.9946

Epoch 20/20  
79/79 [=====] - 19s 239ms/step - loss: 0.0369 - accuracy:  
0.9888

Out[26]:

<keras.callbacks.History at 0x26bd0c53ee0>



In [27]:

```
timesteps = 100
feature = 1000

for i in range(5):
    test_review = " ".join([index_to_word[index] for index in x_test[i]])
    print("=====", str(i + 1), "번째 리뷰=====")
    print(test_review)
    print(" ")

    score = float(model.predict(test_seq[i].reshape(1, timesteps, feature)))

    if (score > 0.5):
        print(str(i + 1), "번째 리뷰는 {:.2f}% 확률로 긍정 리뷰입니다.".format(score * 100))
    else :
        print(str(i + 1), "번째 리뷰는 {:.2f}% 확률로 긍정 리뷰입니다.".format((1 - score) * 100))

    print(" ")
```

===== 1 번째 리뷰=====

? please give this one a miss br br ?? and the rest of the cast ? terrible performances the show is ??? br br i don't know how michael ? could have ? this one on his ? he almost seemed to know this wasn't going to work out and his performance was quite ? so all you ? fans give this a miss

1/1 [=====] - 1s 647ms/step

1 번째 리뷰는 99.69% 확률로 긍정 리뷰입니다.

===== 2 번째 리뷰=====

? this film ? a lot of ? because it ? on ? and character development the plot is very simple and many of the scenes take place on the same set in ?? the ?? character ? but the film ? to a ?? br br the characters create an atmosphere ? with sexual ? and ?? it's very interesting that robert ? directed this ? the style and ? of his other films still the ??? style is ? here and there i think what really makes this film work is the brilliant performance by ?? it's definitely one of her ? characters but she plays it so perfectly and ? that it's scary michael ? does a good job as the ? young man ??? michael ? has a small part the ?? set ? the ? of the story very well in short this movie is a powerful ? of ? sexual ? and ? be ?? up the atmosphere and pay attention to the ? written script br br i ? robert ? this is one of his many films that ? with ?? subject matter this film is ? but it's ? and it's sure to ? a strong emotional ? from the viewer if you want to see an ? film some might even say ? this is worth the time br br unfortunately it's very difficult to find in video ? you may have to buy it off the ?

1/1 [=====] - 0s 62ms/step

2 번째 리뷰는 99.77% 확률로 긍정 리뷰입니다.

===== 3 번째 리뷰=====

? many animation ??? the great ?? of one special ? of the art ? animation which he ? almost single ? and as it happened almost ? as a young man ? was more interested in ? than the cinema but his ? attempt to film two ?? fighting ? to an ?? in film making when he ? he could ?? by ??? and ? them one ? at a time this ?? to the production of ?? classic short the ?? which he made in ? in ? at a time when ? picture animation of all ? was in its ? br br the political ? of the ??? ? to move to ? where one of his first ?? was a dark political ?? known as ? or the ? who wanted a king a ? of black comedy can be found in almost all of films but here it is very dark indeed ? more at ?? who can ? the ?? than children who would most ? find the ? ? i'm middle ? and found it pretty ? myself and indeed ? of the film ? for english ? viewers of the ? were given title ?? with ? and ? in order to help ? the ?? of the ? br br our tale is set in a ? the ?? where the ? are ? with their ? and have called a special ? to see what they can do to ?? they ? to ?? for a king the ? are ?? in this opening sequence it couldn't have been easy to make so many ?? look ?? while ? for his part is ? as a ? white ? guy in the ? who looks like ? rather be taking a ? when ?? them a ? like god who ? them the ?? that this is no ? and ? a different king ?? ? them a ? br br ? with this ? looking new king who ? above them the ?? him with a ? of ??? the ?? forward to hand him the ? to the ? as ??? the ? to ? horror the ??? the ? and then goes on a ??? ? at ? a title ??? ? of the ?? throughout the ? when the now ?? once more ?? for help he ? his ? and ? their ? with ?? the ? of our story ? by a ?? just before he is ? is let well enough alone br br ? the time period when this ? little film was made and ? the fact that it was made by a ?? at the ? of that ??? war it would be easy to see this as a ? about those events ? may or may not have had ?? in mind when he made ? but whatever ? his ? of material the film ? as a ? tale of ??? could be the ??? ? or ? in the ? or any country of any era that ? its ? down and is ? by ? it's a ? film even a ? one in its ? way but its message is no joke

1/1 [=====] - 0s 53ms/step

3 번째 리뷰는 93.55% 확률로 긍정 리뷰입니다.

===== 4 번째 리뷰=====

? i ? love this type of movie however this time i found myself ? to ? the screen sin  
ce i can't do that i will just ? about it this was absolutely ? the things that happ  
en with the dead kids are very cool but the ? people are ? ? i am a ? man pretty big  
and i can ? myself well however i would not do half the stuff the little girl does i  
n this movie also the mother in this movie is ? with her children to the point of ?  
i wish i wasn't so ? about her and her ? because i would have otherwise enjoyed the  
flick what a number she was take my ? and fast forward through everything you see he  
r do until the end also is anyone else getting ? of watching movies that are filmed  
so dark ? one can hardly see what is being filmed as an audience we are ? involved w  
ith the ? on the screen so then why the hell can't we have night ?

1/1 [=====] - 0s 52ms/step

4 번째 리뷰는 99.86% 확률로 긍정 리뷰입니다.

===== 5 번째 리뷰=====

? like some other people ? i'm a die hard ? fan and i loved this game br br this gam  
e starts ? boring but ? me it's worth it as soon as you start your ? the ? are fun a  
nd ? they will ? you ? your mind turns to ? i'm not ? this game is also ? and is ? d  
one br br to keep this ? free i have to keep my ? ? about ? but please try this game  
? be worth it br br story 9 9 action 10 1 it's that good ? 10 attention ? 10 average  
10

1/1 [=====] - 0s 48ms/step

5 번째 리뷰는 94.38% 확률로 긍정 리뷰입니다.

In [28]:

```
loss, accuracy = model.evaluate(test_seq, t_test, verbose=1)
```

```
print("test loss : ", round(loss, 6))
```

```
print("test accuracy : ", round(accuracy * 100, 3), "%")
```

32/32 [=====] - 3s 61ms/step - loss: 1.2912 - accuracy: 0.6

600

test loss : 1.291165

test accuracy : 66.0 %