Al School 6기 6주차

파이썬 기초 – 클래스2

CNN 기초2

CNN 모델을 활용한 객체 분류

Al School 6기 6주차

파이썬 기초 – 클래스2



- 클래스는 객체의 구조와 행동을 정의
- 객체의 클래스는 초기화를 통해 제어
- 클래스는 복잡한 문제를 다루기 쉽도록 만들어줌

airtravel.py

```
class Flight:
pass
```

flight_test.py

```
from Python_basic.airtravel import Flight
#생성한 클래스를 import
```

f = Flight() #클래스 객체 생성 및 변수에 할당 print(type(f))

- 메소드란 클래스 내의 함수
- self: 파이썬 메소드의 첫번째 파라미터명
- 인스턴스 메소드란 객체에서 호출되어질수 있는 함수

airtravel.py

```
class Flight:
def number (self): #메소드 작성
return 'KE081'
```

flight_test.py

```
from Python_basic.airtravel import Flight
#생성한 클래스를 import
```

f = Flight() #클래스 객체 생성 및 변수에 할당, 생성자 print(f.number())

• 생성자와 초기화 airtravel.py

```
class Flight:
    def __init__(self):
        print('init')
        super().__init__()

    def __new__(cls):
        print('new')
        return super().__new__(cls)

    def number(self): #메소드 작성
        return 'KE081'
```

```
from Python_basic.airtravel import Flight
f = Flight()
```

• 초기화 airtravel.py

```
class Flight:
    def __init__(self, number):
        self._number = number

def number(self):
        return self._number
```

```
from Python_basic.airtravel import Flight

f = Flight('KE082')

print(f.number())
print(f._number)
```

• 인스턴스 속성(변수) airtravel.py

```
class Flight:
    def __init__(self, number, passenger_num):
        self.__number = number
        self._passenger_num = passenger_num

def number(self): #메소드 작성
        return self.__number

def add_passenger(self, num):
        self._passenger_num += num
```

```
f1 = Flight('KE082', 0) #클래스 객체 생성 및 변수에 할당
f2 = Flight('KE081', 0)
f1.add_passenger(2)
f2.add_passenger(3)
print(f1._passenger_num)
print(f2._passenger_num)
```

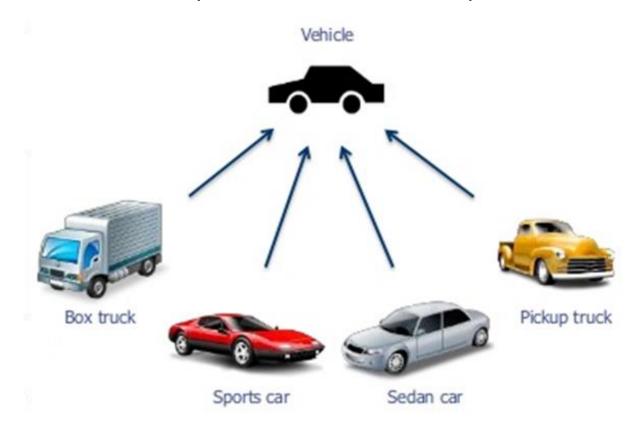
• 클래스 속성

airtravel.py

```
class Flight:
  nation = 'Korea'
.
.
.
```

```
f1 = Flight('KE082', 0) #클래스 객체 생성 및 변수에 할당
f2 = Flight('KE081', 0)
print(f1.nation)
print(f2.nation)
```

- 어떤 클래스를 만들 때 다른 클래스의 기능을 물려받음
- 물려받은 기능을 유지한채로 다른 기능을 추가할 때 사용
- 물려주는 클래스 (Parent class, Super class)
- 물려받는 클래스 (Child class, Sub class)



- 자식 클래스는 이름 옆에 부모클래스의 이름을 입력
- 부모 클래스의 메소드와 속성 모두 사용 가능
- 부모 클래스의 메소드 사용 예제

person.py

```
class Person:
    def greeting(self):
        print('안녕하세요.')

class Student(Person):
    def study(self):
        print('공부하기')
```

```
from Python_basic.person import Student

kim = Student()
kim.greeting()
kim.study()
```

• 부모 클래스의 속성 person.py

```
class Person:
    def __init__(self):
        self.num_arm = 2
        print("Person_init")
        ...

class Student(Person):
    def __init__(self, semester):
        super().__init__()
        print("Student_init")
        self.semester = semester
        ...
```

```
from Python_basic.person import Student

kim = Student(2)
print(kim.num_arm)
print(kim.semester)
```

• 자식 클래스에 __init__이 없으면 super() 생략가능

person.py

```
class Person:
    def __init__(self):
        self.num_arm = 2
        print("Person_init")
        ...
class Student(Person):
    pass
```

```
from Python_basic.person import Student
kim = Student()
print(kim.num_arm)
```

• 메소드 오버라이딩 (Overriding)

person.py

```
class Person:
    def __init__(self):
        self.num_arm = 2
    def greeting(self):
        print('안녕하세요.')
class Student(Person):
    ...
    def greeting(self):
        super().greeting()
        print(f'석사과정 {self.semester}학기생입니다.')
```

```
from Python_basic.person import Student
kim = Student(2)
kim.greeting()
```

• 다중 상속: 여러 부모 클래스로 부터 상속을 받는 것 person.py

```
class Person:
  def __init__(self):
     self.num_arm = 2
  def greeting(self):
     print('안녕하세요.')
class University:
  def credit_show(self):
     print("A")
class Student(Person, University):
```

```
from Python_basic.person import Student
kim = Student(2)
kim.greeting()
kim.credit_show()
```

• 추상 클래스 person.py

```
from abc import *
class StudentBase(metaclass=ABCMeta):
  @abstractmethod
  def study(self):
     pass
  @abstractmethod
  def go_to_school(self):
     pass
class Student(StudentBase):
  def study(self):
     print('공부하기')
  def go_to_school(self):
     print('학교가기')
```

```
from Python_basic.person import Student
kim = Student()
kim.go_to_school()
kim.study()
```

연습문제 (지난 주)

- fourcal.py에서 사칙연산을 수행하는 Calculator 클래스를 만드세요.
- def __init__(self, num1, num2)
- def add(self)

•

•

return result

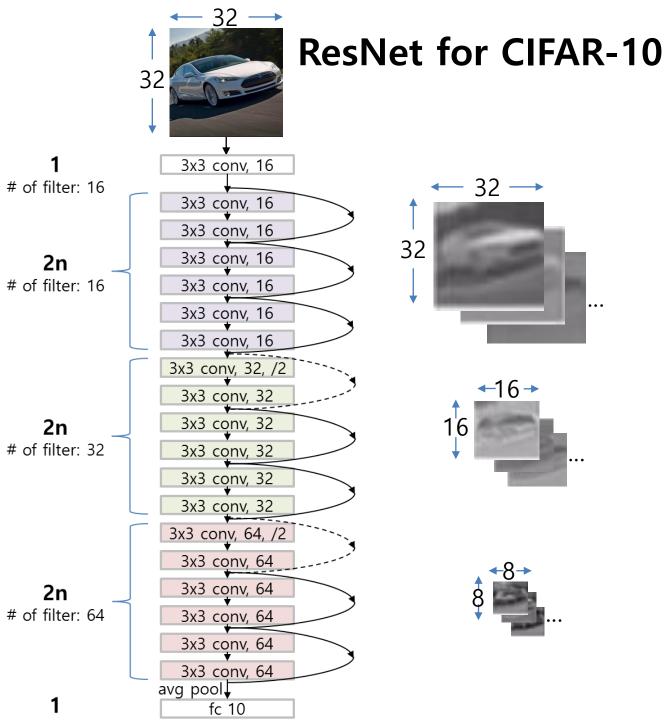
- def sub(self)
- def mul(self)
- def div(self)
- calculator_test.py에서 객체 생성 후 4가지 메소드 사용 결과 출력

연습문제

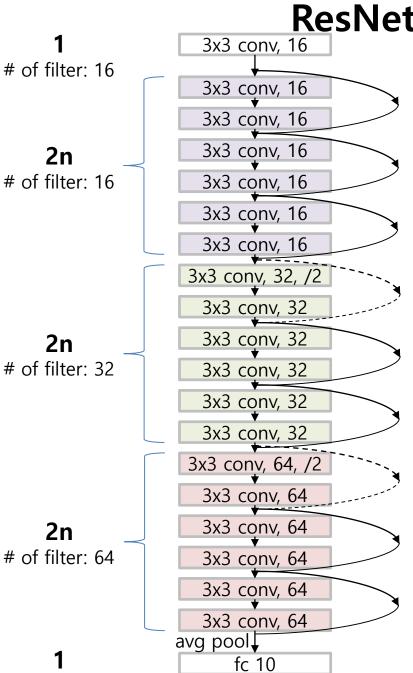
- Calculator 클래스를 상속 받는 ScienCalculator를 만드세요.
- 제곱을 계산하는 def pow(self) 함수를 추가하세요. $num1^{num2}$
- 메소드 오버라이딩을 이용해 div함수가 분모가 0일 때 0을 return하도록 재정의 하세요.
- Calculator_test.py에서 ScienCalculator 객체 생성 후 5가지 메소드 사용 결과 출력
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Al School 6기 6주차

CNN 모델을 활용한 객체 분류



ResNet for CIFAR-10



method			error (%)	
n=3	ResNet	20	0.27M	8.75
n=5	ResNet	32	0.46M	7.51
n=7	ResNet	44	0.66M	7.17
n=9	ResNet	56	0.85M	6.97
n = 18	ResNet	110	1.7M	6.43 (6.61±0.16)
n = 200	ResNet	1202	19.4M	7.93

(6n + 2)

AISchool 6기 저작권:AISchool **Hyperparameters** 3x3 conv, 16 3x3 conv, 16 3x3 conv, 16 3x3 conv, 16 # of filter:16 3x3 conv, 16 3x3 conv, 16 3x3 conv, 16 3x3 conv, 32, /2 3x3 conv, 32 3x3 conv, 32 # of filter:32 3x3 conv, 32 3x3 conv, 32 from tensorflow.keras.datasets.cifar10 import load_data 3x3 conv, 32 3x3 conv, 64, /2 3x3 conv, 64 0보다 작을 때 기울기 3x3 conv, 64 2n tf.flags.DEFINE_float("lr", 0.1, "learning rate (default=0.1)") # of filter:64 3x3 conv, 64 tf.flags.DEFINE_float("lr_decay", 0.1, "learning rate_decay rate(default=0.1)") 3x3 conv, 64 tf.flags.DEFINE_float("12_reg_lambda", 0.0001, "12 regularization lambda (default: 0.0) 3x3 con avg pool 3x3 conv, 64 fc 10

```
tf.flags.DEFINE_float("relu_leakiness", 0.1, "relu leakiness (default: 0.1)")
tf.flags.DEFINE_integer("num_residual_units", 3, "The number of residual_units (default6^{5})"+2
tf.flags.DEFINE_integer("num_classes", 10, "The number of classes (default: 10)")
# Training parameters
                                                                             Unit 개수, 즉 n의 크기
tf.flags.DEFINE_integer("batch_size", 128, "Batch Size (default: 64)")
tf.flags.DEFINE_integer("num_epochs", 100, "Number of training epochs (default: 200)")
tf.flags.DEFINE_integer("evaluate_every", 100, "Evaluate model on dev set after this many steps
(default: 100)")
tf.flags.DEFINE_integer("checkpoint_every", 100, "Save model after this many steps (default: 100)")
tf.flags.DEFINE_integer("num_checkpoints", 3, "Number of checkpoints to store (default: 5)")
# Misc Parameters
tf.flags.DEFINE_boolean("allow_soft_placement", True, "Allow device soft device placement")
tf.flags.DEFINE_boolean("log_device_placement", False, "Log placement of ops on devices")
                                                                                              22
FLAGS = tf.flags.FLAGS
```

resnet_train.py

import CIFAR.data_helpers as dh

from CIFAR.resnet import ResNet

import tensorflow as tf

Model Hyperparameters

import os

import time

import datetime

Data loading & preprocessing

resnet_train.py

```
(x_train_val, y_train_val), (x_test, y_test) = load_data()
x_train, y_train, x_test, y_test, x_val, y_val = dh.shuffle_data(x_train_val, y_train_val, x_test,
y_test)
```

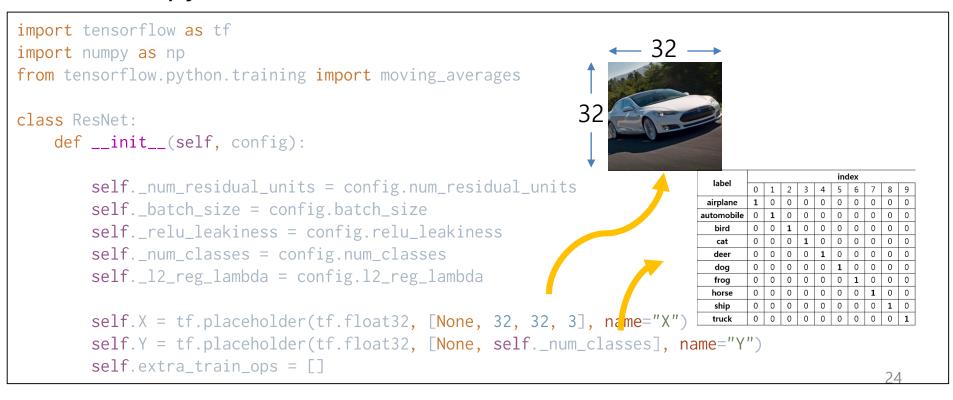
data_helpers.py

```
import numpy as np
                                                                    airplane (0)
                                                                                label
                                                                                              5 6 7
                                                                   automobile (1)
import random
                                                                                     1 0
                                                                                        0
                                                                                          0
                                                                                            0
                                                                                              0 0 0
                                                                                airplane
                                                                               automobile
                                                                                     0 1
                                                                                        0
                                                                                          0
                                                                                              0
                                                                     cat (3)
                                                                     deer (4)
                                                                                       0 0 1 0 0 0
def shuffle_data(x_train_val, y_train_val, x_test, y_test):
                                                                     dog (5)
                                                                                              0
                                                                     frog (6)
                                                                                              1 0
    shuffle_indices = np.random.permutation(np.arange(len(y_train)
                                                                     horse (7)
                                                                                     0 0 0 0 0 0 1
                                                                     ship (8)
    shuffled_x = np.asarray(x_train_val[shuffle_indices])
                                                                                        0
                                                                                          0
                                                                                              0
                                                                                     0 0
                                                                     truck (9)
                                                                                     0 0 0 0 0
                                                                                              0 0
                                                                                ship
    shuffled_v = v_train_val[shuffle_indices]
                                                                                            0 0 0
    val\_sample\_index = -1 * int(0.1 * float(len(y\_train\_val))) original label data?
                                                                                        one-hot-encoded label data
    x_train, x_val = shuffled_x[:val_sample_index], shuffled_x[val_sample_index:]
    v_train, v_val = shuffled_v[:val_sample_index], shuffled_v[:val_sample_index:]
    x_{test} = np.asarray(x_{test})
    0, 0, 0, 1, 0]
    y_train_one_hot = np.squeeze(y_train_one_hot, axis=1) # (45000, 10)
    v_test_one_hot = np.eye(10)[v_test]
    v_test_one_hot = np.squeeze(v_test_one_hot, axis=1)
    y_val_one_hot = np.eye(10)[y_val]
    y_val_one_hot = np.squeeze(y_val_one_hot, axis=1)
    return x_train, y_train_one_hot, x_test, y_test_one_hot, x_val, y_val_one_hot
```

ResNet class & input

resnet_train.py

```
with tf.Graph().as_default():
    session_conf = tf.ConfigProto(
        allow_soft_placement=FLAGS.allow_soft_placement,
        log_device_placement=FLAGS.log_device_placement)
    sess = tf.Session(config=session_conf)
    with sess.as_default():
        resnet = ResNet(FLAGS)
```



Initial convolutional layer

resnet.py

층별 filter 개수 (output channel) 3x3 conv, 16 filters = [16, 16, 32, 64]# of filter: 16 activate_before_residual = [True, False, False] 3x³ conv, 16 3x3 conv, 16 with tf.variable_scope('init'): 3x3 conv, 16 x = self._conv('init_conv', self.X, 3, 3, filters[0], ZN # of filter: 16 3x3 conv, 16 strides=[1, 1, 1, 1]) <u>최초 convolutional laver</u> 3x3 conv, 16 3x3 conv, 16 3x3 conv, 32, /2 3x3 conv, 32 3x3 conv, 32 2n # of filter: 32 3x3 conv, 32 3x3 conv, 32 3x3 conv, 32 3x3 conv, 64, /2 3x3 conv, 64 3x3 conv, 64 2n # of filter: 64 3x3 conv, 64 3x3 conv, 64 3x3 conv, 64 avg pool↓ fc 10 25

Residual units

resnet.py 3x3 conv, 16 n = 3일 때 3x3 conv, 16 with tf.variable_scope('unit_1_0'): $x = self._residual(x, filters[0], filters[1], activate_before_residual[0], str$ 3x3 conv, 16 for i in range(1, self._num_residual_units): 3x3 conv, 16 with tf.variable_scope('unit_1_%d' % i): unit 1 1~2 3x3 conv, 16 $x = self._residual(x, filters[1], filters[1], strides=[1, 1, 1, 1])$ 3x3 conv. 16 3x3 conv, 16 with tf.variable_scope('unit_2_0'): 3x3 conv, 32, /2 x = self._residual(x, filters[1], filters[2], activate_before_residual[1] for i in range(1, self._num_residual_units): 3x3 conv, 32 with tf.variable_scope('unit_2_%d' % i): 3x3 conv, 32 $x = self._residual(x, filters[2], filters[2], strides=[1, 1, 1unit)2 1~2$ 3x3 conv, 32 3x3 conv, 32 with tf.variable_scope('unit_3_0'): x = self._residual(x, filters[2], filters[3], activate_before_residual[2], str_ues_1 + 222 3x3 conv, 64, /2 for i in range(1, self._num_residual_units): unit 3 0 3x3 conv. 64 with tf.variable_scope('unit_3_%d' % i): $x = self._residual(x, filters[3], filters[3], strides=[1, 1, 1, 1])$ 3x3 conv, 64 unit_3_1~2 3x3 conv, 64 with tf.variable_scope('unit_last'): 3x3 conv. 64 x = self._batch_norm('final_bn', x) 3x3 conv. 64

x = self._relu(x, self._relu_leakiness)

x = self._global_avg_pool(x)

unit last

avg pool↓

fc 10

Fully connected layer, weight decay

3x3 conv, 16 resnet.py 3x3 conv, 16 with tf.variable_scope('logit'): 3x3 conv, 16 logits = self._fully_connected(x, self._num_classes) 3x3 conv, 16 self.predictions = tf.nn.softmax(logits) 3x3 conv, 16 self.predictions = tf.argmax(self.predictions, 1, name="predictions") 3x3 conv, 16 with tf.variable_scope('loss'): 3x3 conv, 16 xent = tf.nn.softmax_cross_entropy_with_logit (logits=logits, labels=self.Y) 3x3 conv, 32, /2 self.loss = tf.reduce_mean(xent, name='xent') 3x3 conv, 32 self.loss += self._decay() 3x3 conv, 32 3x3 conv, 32 with tf.name_scope("accuracy"): correct_predictions = tf.equal(self.predictions tf.argmax(self.Y, 1)) 3x3 conv, 32 self.accuracy = tf.reduce_mean(tf.cast(correct_pledictions, "float"), name="a 3x3 conv, 32" 3x3 conv, 64, /2 3x3 conv, 64 3x3 conv, 64 3x3 conv, 64 10개의 label 중 하나로 분류하기 위<mark>한</mark> 3x3 conv, 64 **Fully connected layer** 3x3 conv, 64 avg pool↓ fc 10

Residual

```
def _residual(self, x, in_filter, out_filter, activate_before_residual=False, strides=[1, 1, 1,
17):
    if activate_before_residual:
                                                                         orig_x x
        with tf.variable_scope('common_activation'):
                                                                               weight layer
                                                                        sub1
            x = self._batch_norm('init_bn', x)
            x = self._relu(x, self._relu_leakiness)
                                                                     \mathcal{F}(\mathbf{x})
                                                                                     relu
                                                                                                    \mathbf{X}
            orig_x = x
                                                                        sub2
                                                                               weight layer
                                                                                                  identity
    else:
        with tf.variable_scope('residual_activation'):
                                                            \textbf{x+=orig\_x}~\mathcal{F}(\mathbf{x}) + \mathbf{x}
            orig_x = x
                                                                                    ↓ relu
            x = self._batch_norm('init_bn', x)
            x = self._relu(x, self._relu_leakiness)
                                                                                 3x3 conv, 32, /2
    with tf.variable_scope('sub1'):
                                                                                  3x3 conv, 32
        x = self._conv('conv1', x, 3, in_filter, out_filter, strides)
    with tf.variable_scope('sub2'):
                                                                                  3x3 conv, 32
        x = self._batch_norm('bn2', x)
                                                                                  3x3 conv, 32
        x = self._relu(x, self._relu_leakiness)
                                                                                  3x3 conv, 32
       x = self. \_conv('conv2', x, 3, out\_filter, out\_filter, [1, 1, 1, 1]
    with tf.variable_scope('sub_add'):
                                                                                  3x3 conv, 32
        if in_filter != out_filter:
            orig_x = tf.nn.avg_pool(orig_x, strides, 'VALID')
            orig_x = tf.pad(orig_x, [[0, 0], [0, 0], [0, 0], [(out_filter_-in_filter) / 변경됨에 따라
(out_filter - in_filter) // 2]])
                                                                     동일 한 크기의 feature map간에
        x += orig_x
                                                                  Skip connection이 이루어지지 않을 때
    tf.logging.debug('image after unit %s', x.get_shape())
    return x
```

// scale and shift

 $y_i \leftarrow \gamma \widehat{x}_i + \beta \equiv BN_{\gamma,\beta}(x_i)$

Batch normalization

```
def _relu(self, x, leakiness=0.0):
     return tf.where(tf.less(x, 0.0), leakiness * x, x, name='leaky_relu')
def _batch_norm(self, name, x):
    with tf.variable_scope(name):
         params_shape = [x.get_shape()[-1]]
         beta = tf.get_variable('beta', params_shape, tf.float32,
initializer=tf.constant_initializer(0.0, tf.float32))
         gamma = tf.get_variable('gamma', params_shape, tf.float32,
initializer=tf.constant_initializer(1.0, tf.float32))
         mean, variance = tf.nn.moments(x, [0, 1, 2], name='moments')
         moving_mean = tf.get_variable('moving_mean', params_shape, tf.float32,
               initializer=tf.constant_initializer(0.0, tf.float32), trainable=False)
         moving_variance = tf.get_variable('moving_variance', params_shape, tf.float32,
               initializer=tf.constant_initializer(1.0, tf.float32), trainable=False)
          self.extra_train_ops.append(moving_averages.assign_moving_
                                                                                     Input: Values of x over a mini-batch: \mathcal{B} = \{x_1, y_n\}:
          self.extra_train_ops.append(moving_averages.assign_moving_
                                                                                           Parameters to be learned: \gamma, \beta
                                                                                     Output: \{y_i = BN_{\gamma,\beta}(x_i)\}
(0.9)
                                                                                      \mu_{\mathcal{B}} \leftarrow \frac{1}{m} \sum_{i=1}^{m} x_i
                                                                                                                 // mini-batch mean
         y = tf.nn.batch_normalization(x, mean, variance, beta, gam
         y.set_shape(x.get_shape())
                                                                                       \sigma_{\mathcal{B}}^2 \leftarrow \frac{1}{m} \sum_{i=1}^m (x_i - \mu_{\mathcal{B}})^2
                                                                                                              // mini-batch variance
         return y
                                                                                       \widehat{x}_i \leftarrow \frac{x_i - \mu_B}{\sqrt{\sigma_B^2 + \epsilon}}
                                                                                                                      // normalize
```

Fully connected layer, weight decay

```
def _fully_connected(self, x, out_dim):
    dim = tf.reduce_prod(x.get_shape()[1:]).eval()
    x = tf.reshape(x, [-1, dim])
    w = tf.get_variable('DW', [dim, out_dim],
        initializer=tf.uniform_unit_scaling_initializer(factor=1.0))
    b = tf.get_variable('biases', [out_dim], initializer=tf.constant_initializer())
    return tf.nn.xw_plus_b(x, w, b)
```

```
def _global_avg_pool(self, x):
    assert x.get_shape().ndims == 4
    return tf.reduce_mean(x, [1, 2])
```

```
def _decay(self):
    """L2 weight decay loss."""
    costs = []
    for var in tf.trainable_variables():
        if var.op.name.find(r'DW') > 0:
            costs.append(tf.nn.l2_loss(var))

    return tf.multiply(self._l2_reg_lambda, tf.add_n(costs))
```

Optimizer

```
with tf.Graph().as_default():
    session_conf = tf.ConfigProto(
      allow_soft_placement=FLAGS.allow_soft_placement,
      log_device_placement=FLAGS.log_device_placement)
    sess = tf.Session(config=session_conf)
   with sess.as_default():
        resnet = ResNet(FLAGS)
        # Define Training procedure
        global_step = tf.Variable(0, name="global_step", trainable=False)
       decayed_lr = tf.train.exponential_decay(FLAGS.lr, global_step, 24000, FLAGS.lr_decay,
staircase=True)
        optimizer = tf.train.MomentumOptimizer(learning_rate=decayed_lr, momentum=0.9)
        grads_and_vars = optimizer.compute_gradients(resnet.loss)
        train_op = optimizer.apply_gradients(grads_and_vars, global_step=global_step)
        train_ops = [train_op] + resnet.extra_train_ops
        train_ops = tf.group(*train_ops)
        # Output directory for models and summaries
        timestamp = str(int(time.time()))
        out_dir = os.path.abspath(os.path.join(os.path.curdir, "runs", timestamp))
        print("Writing to {}\n".format(out_dir))
```

Summary, checkpoint

```
# Summaries for loss and accuracy
loss_summary = tf.summary.scalar("loss", resnet.loss)
acc_summary = tf.summary.scalar("accuracy", resnet.accuracy)
# Train Summaries
train_summary_op = tf.summary.merge([loss_summary, acc_summary])
train_summary_dir = os.path.join(out_dir, "summaries", "train")
train_summary_writer = tf.summary.FileWriter(train_summary_dir, sess.graph)
# Dev summaries
dev_summary_op = tf.summary.merge([loss_summary, acc_summary])
dev_summary_dir = os.path.join(out_dir, "summaries", "dev")
dev_summary_writer = tf.summary.FileWriter(dev_summary_dir, sess.graph)
# Checkpoint directory. Tensorflow assumes this directory already exists so we need to create it
checkpoint_dir = os.path.abspath(os.path.join(out_dir, "checkpoints"))
checkpoint_prefix = os.path.join(checkpoint_dir, "model")
if not os.path.exists(checkpoint_dir):
    os.makedirs(checkpoint_dir)
saver = tf.train.Saver(tf.global_variables(), max_to_keep=FLAGS.num_checkpoints)
```

Train & dev step

```
sess.run(tf.global_variables_initializer())
def train_step(x_batch, y_batch):
    feed_dict = {
     resnet.X: x_batch,
      resnet.Y: y_batch
    _, step, lr, summaries, loss, accuracy = sess.run(
       [train_ops, global_step, decayed_lr, train_summary_op, resnet.loss, resnet.accuracy],
       feed_dict)
    time_str = datetime.datetime.now().isoformat()
    print("{}: step {}, lr {}, loss {:g}, acc {:g}".format(time_str, step, lr, loss, accuracy))
    train_summary_writer.add_summary(summaries, step)
def dev_step(x_batch, y_batch, writer=None):
feed_dict = {
      resnet.X: x_batch,
      resnet.Y: y_batch
    step, summaries, loss, accuracy = sess.run(
       [global_step, dev_summary_op, resnet.loss, resnet.accuracy],
       feed_dict)
    time_str = datetime.datetime.now().isoformat()
    print("{}: step {}, loss {:g}, acc {:g}".format(time_str, step, loss, accuracy))
    if writer:
       writer.add_summary(summaries, step)
                                                                                             33
    return accuracy
```

Batch, Training

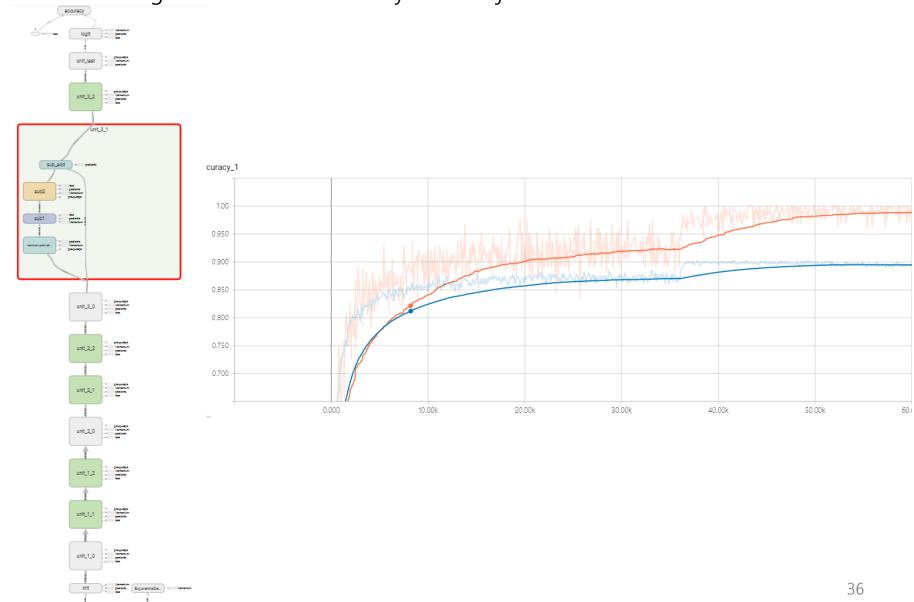
Batch

data_helpers.py

```
def batch_iter(x, y , batch_size, num_epochs, shuffle=True):
   num_batches_per_epoch = int((len(x) - 1) / batch_size) + 1
   for epoch in range(num_epochs):
       shuffle_indices = np.random.permutation(np.arange(len(x)))
       shuffled_x = x[shuffle_indices]
       shuffled_y = y[shuffle_indices]
       for batch_num in range(num_batches_per_epoch):
                                                                            학습 데이터를 늘리기 위해
           start_index = batch_num * batch_size
                                                                      좌우 대칭 및 사진의 일부영역 잘라내기
           end_index = min((batch_num + 1) * batch_size, len(y))
           yield list(zip(data_augmentation(shuffled_x[start_index:end_index], 4),
shuffled_v[start_index:end_index]))
def data_augmentation (x_batch, padding=None):
   for i in range(len(x_batch)):
       if bool(random.getrandbits(1)):
           x_batch[i] = np.fliplr(x_batch[i])
   oshape = np.shape(x_batch[0])
   if padding:
       oshape = (oshape[0] + 2 * padding, oshape[1] + 2 * padding)
   new batch = []
   npad = ((padding, padding), (padding, padding), (0, 0))
   for i in range(len(x_batch)):
       new_batch.append(x_batch[i])
       if padding:
           new_batch[i] = np.lib.pad(x_batch[i], pad_width=npad, node=
                                                                                 constant_valu
       nh = random.randint(0, oshape[0] - 32)
       nw = random.randint(0, oshape[1] - 32)
       new_batch[i] = new_batch[i][nh:nh + 32, nw:nw + 32]
   return new_batch
```

Tensorboard

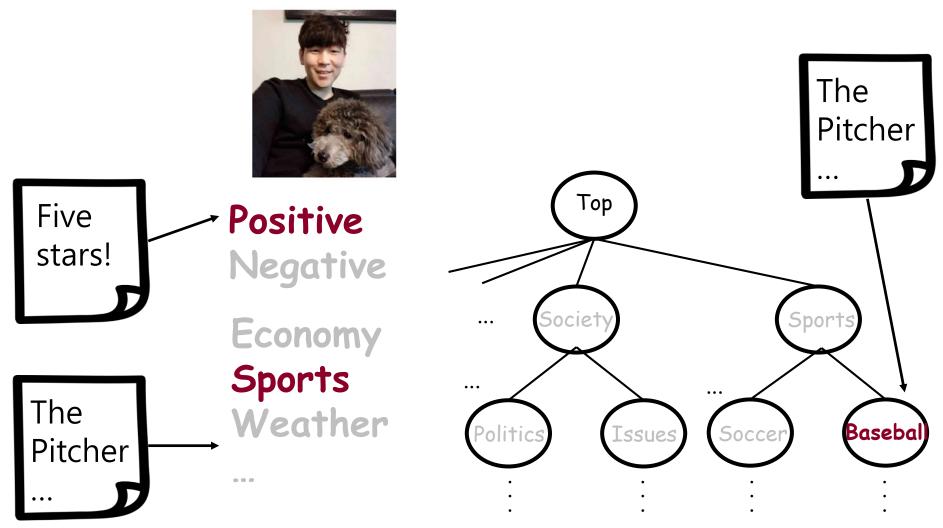
tensorboard --logdir=C:₩Users₩82102₩PycharmProjects₩aischool₩CIFAR₩runs₩1580556863



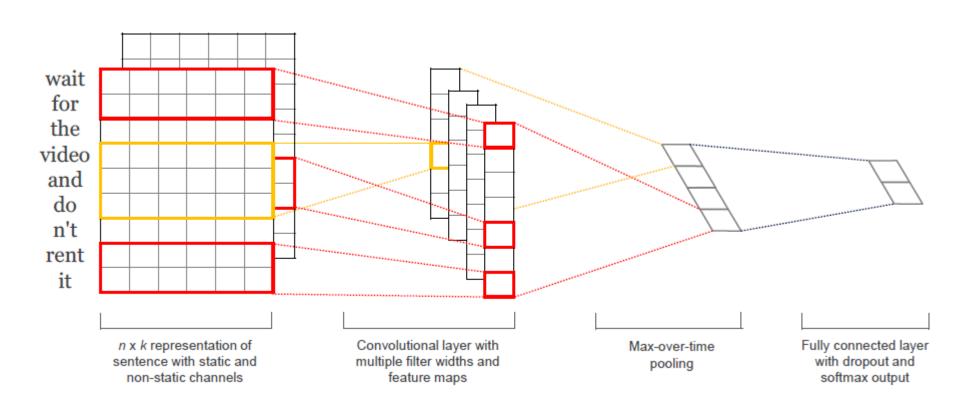
Al School 6기 6주차

CNN 모델을 활용한 영화리뷰 평점 예측

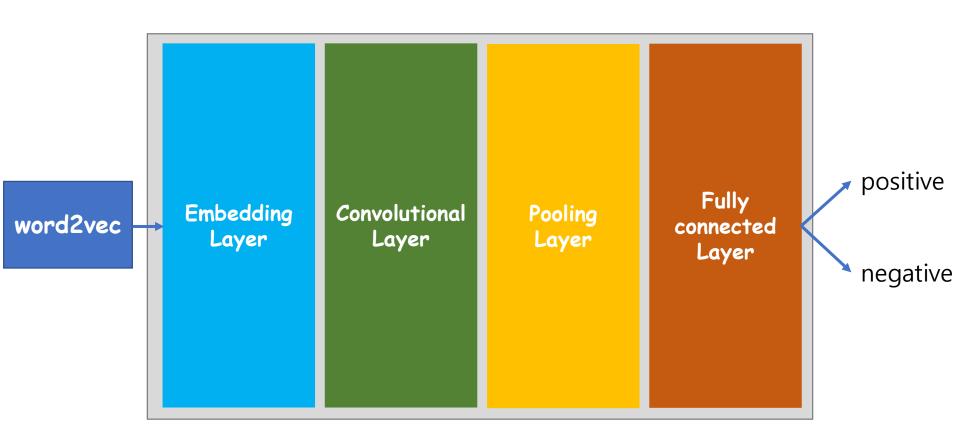
CNN for sentence classification [Kim et al., 2014]



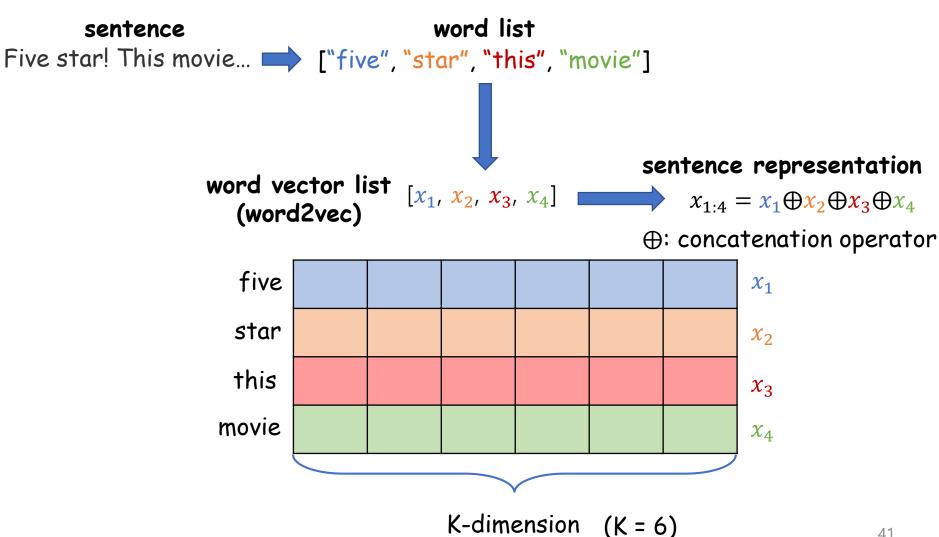
Model overview



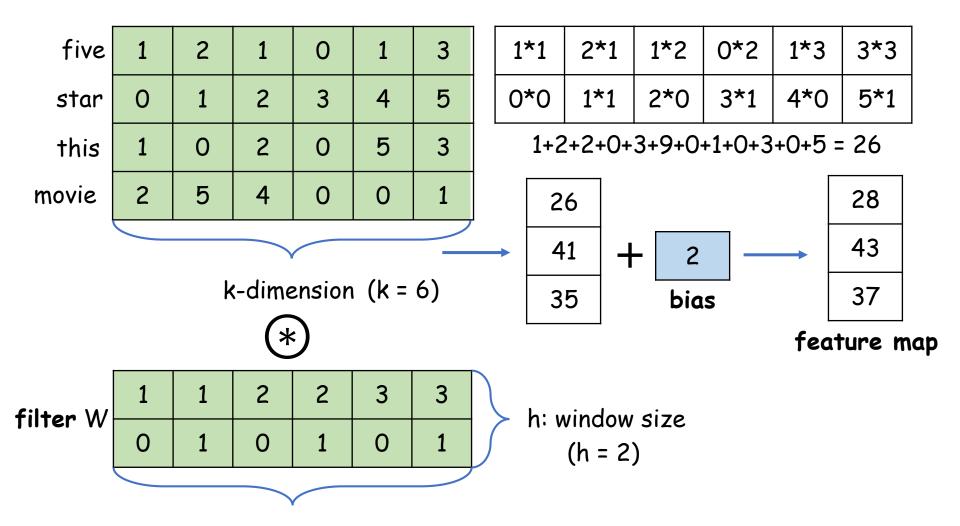
Model overview



Word2vec & Embedding layer



Convolutional layer



Pooling layer

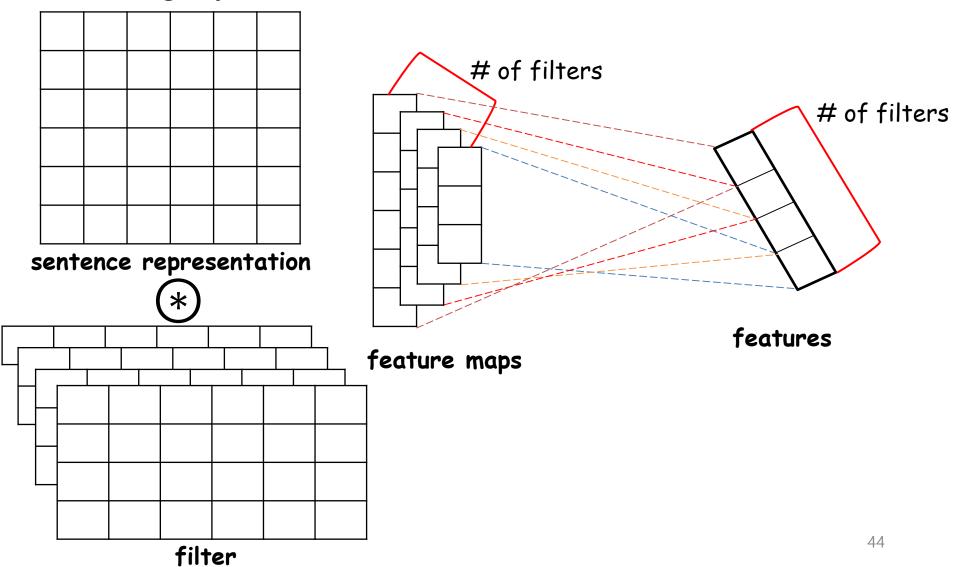
five	1	2	1	0	1	3
star	0	1	2	3	4	5
this	1	0	2	0	5	3
movie	2	5	4	0	0	1

	28	Max-over-	
	43	time pooling	43
	37		 feature
fec	ature r	nap	

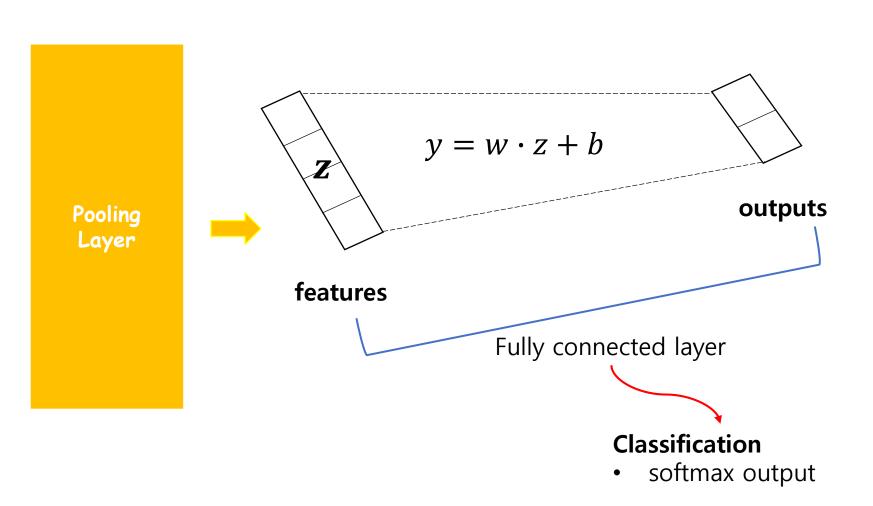
filter W 1 1 2 2 3 3 3 0 1 0 1 0 1

One filter -> One feature
Multiple filter -> Multiple feature

Pooling layer



Fully connected layer



CNN 분류기 학습

• 학습 데이터 다운로드

PC > 새 볼륨 (D:) > Anaconda > workspace > CNN4TC > data				
이름	수정한 날짜	유형 크기		
rt-polaritydata	2019-04-27 오전 8:21	파일 쫄더		
test	2019-05-26 오전 2:16	파일 폴더		
train	2019-05-25 오후 8:48	파일 폴더		

CNN 분류기 학습

Import

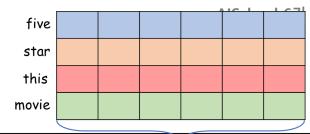
train.py

```
import tensorflow as tf
import numpy as np
import os
import time
import datetime
import CNN4TC.data_helpers as dh
from CNN4TC.text_cnn import TextCNN
from tensorflow.contrib import learn
```

data_helpers.py

```
Jimport numpy as np
import re
Jimport glob
```

Hyperparameters



train.py

```
# Data loading params
                                                                                              K-dimension
tf.flags.DEFINE_float("dev_sample_percentage", .1, "Percentage of the training data to use for validation")
tf.flags.DEFINE_string("imdb_pos_data_file", "./data/train/pos/*", "Data source for the positive data.")
tf.flags.DEFINE_string("imdb_neg_data_file", "./data/train/neg/*", "Data source for the regative data.")
# Model Hyperparameters
# Model Hyperparameters

단어를 표현하는 벡터의 크기
tf.flags.DEFINE_integer("embedding_dim", 128, "Dimensionality of word embedding (default: 128)")
tf.flags.DEFINE_string("filter_sizes", "2,3,4", "Comma_separated filter sizes (N-gram) (default: '3,4,5')")
tf.flags.DEFINE_integer("num_filters", 128, "Number of filters per filter size (default: 128)")
tf.flags.DEFINE_float("lr", 5e-4, 'learning rate (default: 0.001)")
                                                                                                     Filter의 높이
tf.flags.DEFINE_float("dropout_keep_prob", 0.6, "Dropout keep probability (default: 0.5)")
                                                                                                          or
tf.flags.DEFINE float("I2 reg lambda", 0.000000001, "L2 regularization lambda (default: 0.0)
                                                                                                    Window size
# Training parameters
                                                                                                          or
tf.flags.DEFINE_integer("batch_size", 64, "Batch Size (default: 64)")
                                                                                                       N-gram
tf.flags.DEFINE_integer("num_epochs", 20, "Number of training epochs (default: 200)")
tf.flags.DEFINE integer("evaluate every", 100, "Evaluate model on dev set after this many steps (default:
100)")
tf.flags.DEFINE_integer("checkpoint_every", 100, "Save model after this many steps (iterations) (default: 100)")
tf.flags.DEFINE_integer("num_checkp\ints", 3, "Number of checkpoints to store (default: 5)")
# Misc Parameters
tf.flags.DEFINE_boolean("allow_soft_placement", True, "Allow device soft device placement")
tf.flags.DEFINE_boolean("log_device_placement", False, "Log placement of ops on devices")
FLAGS = tf.flags.FLAGS
                            Filter 종류별 개수
```

filter V

h: window size

(h = 2)

Data loading & preprocessing

train.py

```
# Load data
print("Loading data...")
x text, y = dh.load imdb data and labels(FLAGS.imdb pos data file, FLAGS.imdb neg data file)
                                                                         모델이 처리할 수 있는
# Build vocabulary
                                                                            문장의 최대 길이
max_document_length = max([len(x.split(" ")) for x in x_text])
print("max_document_length: ", max_document_length) #298
                                                                                              video
                                                                                              and
vocab_processor = learn.preprocessing.VocabularyProcessor(max_document_length)
x = np.array(list(vocab_processor.fit_transform(x_text)))
                      five star this movie is good--> [8 379 3 47574 2 45 0 0 0 0 0 0 0 0 0 0 ...]
np.random.seed(10)
shuffle_indices = np.random.permutation(np.arange(len(y)))
x \text{ shuffled} = x[\text{shuffle indices}]
y_shuffled = y[shuffle_indices]
# Split train/test set
dev_sample_index = -1 * int(FLAGS.dev_sample_percentage * float(len(y)))
x train, x dev = x shuffled[:dev sample index], x shuffled[dev sample index:]
y train, y dev = y shuffled[:dev sample index], y shuffled[dev sample index:]
del x, y, x_shuffled, y_shuffled
print("Vocabulary Size: {:d}".format(len(vocab processor.vocabulary )))
```

print("Train/Dev split: {:d}/{:d}".format(len(y train), len(y dev)))

IMDB data loading

data_helpers.py

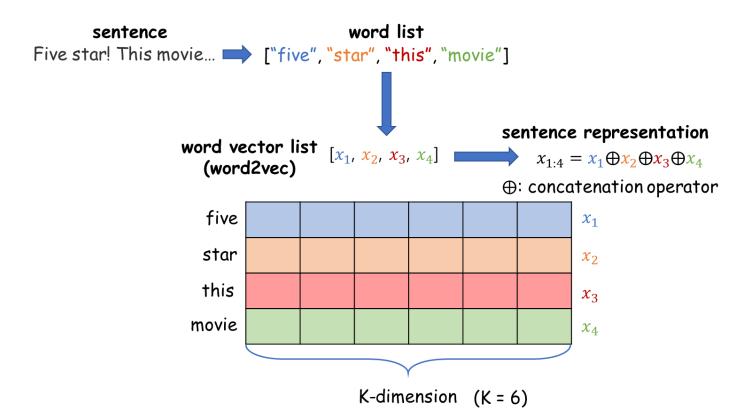
```
def load_imdb_data_and_labels(pos_file, neg_file):
   # Load data from files
                                                                         list = [[1,2,3], [3,6,9], [2,4,6]]
   pos_list = glob.glob(pos_file) #load file list
                                                                          matrix = numpy.array(list)
   pos final = []# sentence list
                                                                          matrix = numpy.zeros((3,3))
   for pos in pos list:
      x_text = list(open(pos, "r", encoding='UTF8').readlines())
      x text = [clean str(sent) for sent in x text]
                                                                          matrix =
      pos_final = pos_final + x_text
                                                                          numpy.random.rand(3,3)
   neg_list = glob.glob(neg_file)
                                                                          matrix.shape
   neg final = []
   for neg in neg_list:
      x_text = list(open(neg, "r", encoding='UTF8').readlines())
      x_text = [clean_str(sent) for sent in x_text]
      neg final = neg final + x text
   positive_labels = [[0, 1] for _ in pos_final] #[[0,1], [0,1], [0,1], [0,1], [0,1], ...]
   negative_labels = [[1, 0] for _ in neg_final] #[[1,0], [1,0], [1,0], [1,0], [1,0],...]
   y = \text{np.concatenate}([\text{positive\_labels}, \text{negative\_labels}], 0) ##[[0,1], [0,1], [0,1], [0,1], [0,1], [0,1], [1,0]
   x final = pos final + neg final
   return [x_final, y]
```

TextCNN class & input

```
import tensorflow as tf
class TextCNN(object):
   def init (self, sequence length, num classes, vocab size, embedding size, filter sizes, num filters,
12 reg lambda=0.0):
      # Placeholders for input, output and dropout
      self.input_x = tf.placeholder(tf.int32, [None, sequence_length], name="input_x")
      self.input_y = tf.placeholder(tf.float32, [None, num_classes], name="input_y")#pos: [0 1] neg: [1 0]
      self.dropout keep prob = tf.placeholder(tf.float32, name="dropout keep prob") #training: 0.5 #test:
1.0
      # Keeping track of 12 regularization loss (optional)
      12 loss = tf.constant(0.0)
```

Embedding layer

```
# Embedding layer
with tf.device('/gpu:0'), tf.name_scope("embedding"):
    self.W = tf.Variable(tf.random_uniform([vocab_size, embedding_size], -1.0, 1.0), name="W")
    self.embedded_chars = tf.nn.embedding_lookup(self.W, self.input_x) #[batch, embedding_dim_,
sequenth_length]
    self.embedded_chars_expanded = tf.expand_dims(self.embedded_chars, -1)#[batch, embedding_dim_,
sequenth_length, 1]
```



Convolutional layer

```
# Create a convolution + maxpool layer for each filter size
                                                                            num_filter
pooled outputs = []
for i, filter size in enumerate(filter sizes):
   with tf.name scope("conv-maxpool-%s" % filter size):
                                                                           filter size
      # Convolution Layer
      filter shape = [filter size, embedding size, 1, num filters]
                                                                                             embedding size
      W = tf.Variable(tf.truncated_normal(filter_shape, stddev=0.1), name="W")
      b = tf.Variable(tf.constant(0.1, shape=[num filters]), name="b")
      conv = tf.nn.conv2d(
         self.embedded_chars_expanded
                                                               five
                                                                              0
                                                                                     3
         W____
                                                                                             (N-F)/S + 1
         strides=[1, 1, 1, 1],
                                                                              0
                                                              this
         padding="VALID",
                                                                   2
                                                             movie
                                                                              0
                                                                                 0
                                                                                                      Max-over-
         name="conv")
                                                                                                      time pooling
                                                                                    (N-F)/S + 1
                                                                                                               43
      # Apply nonlinearity
                                                                                                             feature
      h = tf.nn.relu(tf.nn.bias add(conv, b), name=\"relu")
                                                                                              feature map
      # Maxpooling over the outputs
                                                                                              One filter -> One feature
      pooled = tf.nn.max pool(
                                                            filter W
                                                                                           Multiple filter -> Multiple feature
         ksize=[1, sequence_length - filter_size + 1, 1, 1], #mnist [1,2,2,1]
         strides=[1, 1, 1, 1],
         padding='VALID',
         name="pool")
      pooled outputs.append(pooled)
```

Fully connected layer

```
# Combine all the pooled features
                                                                                       # of filters
num filters_total = num_filters * len(filter_sizes)
                                                                                                               # of filters
self.h_pool = tf.concat(pooled_outputs, 3)
self.h_pool_flat = tf.reshape(self.h_pool, [-1, num_filters_total])
with tf.name scope("dropout"):
   self.h_drop = tf.nn.dropout(self.h_pool_flat, self.dropout_keep_prob)
                                                                                                         features
                                                                                feature maps
with tf.name_scope("output"):
   W = tf.get variable(
      "W".
      shape=[num_filters_total, num_classes],
                                                                                                    \mathbf{w} \cdot \mathbf{z} + \mathbf{b}
      initializer=tf.contrib.layers.xavier_initializer())
                                                                                                                   outputs
   b = tf.Variable(tf.constant(0.1, shape=[num_classes]), name="b")
   12 loss += tf.nn.12 loss(W)
                                                                                         features
   12 loss += tf.nn.l2_loss(b)
                                                                                                     Fully connected layer
   self.scores = tf.nn.xw_plus_b(self.h_drop, W, b, name="scores")
   self.predictions = tf.argmax(self.scores, 1, name="predictions")
with tf.name scope("loss"):
   losses = tf.nn.softmax_cross_entropy_with_logits(logits=self.scores, labels=self.input_y)
   self.loss = tf.reduce mean(losses) + 12 reg lambda * 12 loss
with tf.name_scope("accuracy"):
   correct_predictions = tf.equal(self.predictions, tf.argmax(self.input_y, 1))
   self.accuracy = tf.reduce_mean(tf.cast(correct_predictions, "float"), name="accuracy")
                                                                                                                54
```

Optimizer

```
with tf.Graph().as default():
  session_conf = tf.ConfigProto(
    allow soft placement=FLAGS.allow soft placement,
    log_device_placement=FLAGS.log_device_placement)
   sess = tf.Session(config=session_conf)
   with sess.as default():
      cnn = TextCNN(
         sequence_length=x_train.shape[1],
         num classes=y train.shape[1],
         vocab_size=len(vocab_processor.vocabulary_),
         embedding_size=FLAGS.embedding_dim,
         filter_sizes=list(map(int, FLAGS.filter_sizes.split(","))),
         num filters=FLAGS.num filters,
         I2_reg_lambda=FLAGS.I2_reg_lambda)
     # Define Training procedure
      global_step = tf.Variable(0, name="global_step", trainable=False)
      optimizer = tf.train.AdamOptimizer(FLAGS.lr)
      #optimizer = tf.train.AdagradOptimizer(FLAGS.lr)
      grads_and_vars = optimizer.compute_gradients(cnn.loss)
      train_op = optimizer.apply_gradients(grads_and_vars, global_step=global_step)
```

Summaries

```
# Keep track of gradient values and sparsity (optional)
grad summaries = []
for g, v in grads_and_vars:
   if g is not None:
      grad_hist_summary = tf.summary.histogram("{}/grad/hist".format(v.name), g)
     sparsity_summary = tf.summary.scalar("{}/grad/sparsity".format(v.name), tf.nn.zero_fraction(g))
     grad_summaries.append(grad_hist_summary)
     grad_summaries.append(sparsity_summary)
grad_summaries_merged = tf.summary.merge(grad_summaries)
# Output directory for models and summaries
timestamp = str(int(time.time()))
out_dir = os.path.abspath(os.path.join(os.path.curdir, "runs", timestamp))
# Summaries for loss and accuracy
loss_summary = tf.summary.scalar("loss", cnn.loss)
acc_summary = tf.summary.scalar("accuracy", cnn.accuracy)
# Train Summaries
train_summary_op = tf.summary.merge([loss_summary, acc_summary, grad_summaries_merged])
train_summary_dir = os.path.join(out_dir, "summaries", "train")
train_summary_writer = tf.summary.FileWriter(train_summary_dir, sess.graph)
# Dev summaries
dev_summary_op = tf.summary.merge([loss_summary, acc_summary])
dev_summary_dir = os.path.join(out_dir, "summaries", "dev")
dev_summary_writer = tf.summary.FileWriter(dev_summary_dir, sess.graph)
```

Checkpoint, Save vocabulary

Train & dev step

```
def train_step(x_batch, y_batch):
   feed dict = {
    cnn.input x: x batch,
    cnn.input v: v batch,
    cnn.dropout keep prob: FLAGS.dropout keep prob
   _, step, summaries, loss, accuracy = sess.run(
      [train_op, global_step, train_summary_op, cnn.loss, cnn.accuracy],
      feed dict)
   time str = datetime.datetime.now().isoformat()
   print("{}: step {}, loss {:g}, acc {:g}".format(time_str, step, loss, accuracy))
   train_summary_writer.add_summary(summaries, step)
def dev_step(x_batch, y_batch, writer=None):
   feed dict = {
    cnn.input x: x batch,
    cnn.input_y: y_batch,
    cnn.dropout_keep_prob: 1.0
   step, summaries, loss, accuracy = sess.run(
      [global_step, dev_summary_op, cnn.loss, cnn.accuracy],
      feed dict)
   time str = datetime.datetime.now().isoformat()
   print("{}: step {}, loss {:g}, acc {:g}".format(time_str, step, loss, accuracy))
   if writer:
      writer.add summary(summaries, step)
                                                                                                         58
   return accuracy
```

Batch, Training

train.py

```
batches = dh.batch_iter(list(zip(x_train, y_train)), FLAGS.batch_size, FLAGS.num_epochs)
max = 0
for batch in batches:
    x_batch, y_batch = zip(*batch)
    train_step(x_batch, y_batch)
    current_step = tf.train.global_step(sess, global_step)
    if current_step % FLAGS.evaluate_every == 0:
        accuracy = dev_step(x_dev, y_dev, writer=dev_summary_writer)
        if accuracy > max:
            max = accuracy
            path = saver.save(sess, checkpoint_prefix, global_step=current_step)
            print("Saved model checkpoint to {}\format(path))
```

data_helpers.py

```
def batch_iter(data, batch_size, num_epochs, shuffle=True):
    data = np.array(data)
    data_size = len(data)
    num_batches_per_epoch = int((len(data)-1)/batch_size) + 1
    for epoch in range(num_epochs):
        if shuffle:
            shuffle_indices = np.random.permutation(np.arange(data_size))
            shuffled_data = data[shuffle_indices]
        else: shuffled_data = data
        for batch_num in range(num_batches_per_epoch):
            start_index = batch_num * batch_size
            end_index = min((batch_num + 1) * batch_size, data_size)
            yield shuffled_data[start_index:end_index]
```

CNN 분류기 학습

• ./run/생성시간/checkpoints- 학습 모델 확인

내 PC > 새 볼륨 (D:) > Anaconda > workspace > CNN4TC > runs > 1558802980 > checkpoints				
이를	^	수정한 날짜	유형	크기
	checkpoint	2019-05-26 오전	파일	1KB
	model-700.data-00000-of-00001	2019-05-26 오전	DATA-00000-OF	73,682KB
	model-700.index	2019-05-26 오전	INDEX 파일	2KB
	model-700.meta	2019-05-26 오전	META 파일	121KB
	model-800.data-00000-of-00001	2019-05-26 오전	DATA-00000-OF	73,682KB
	model-800.index	2019-05-26 오전	INDEX 파일	2KB
	model-800.meta	2019-05-26 오전	META 파일	121KB
	model-900.data-00000-of-00001	2019-05-26 오전	DATA-00000-OF	73,682KB
	model-900.index	2019-05-26 오전	INDEX 파일	2KB
	model-900.meta	2019-05-26 오전	META 파일	121KB
	model-1000.data-00000-of-00001	2019-05-26 오전	DATA-00000-OF	73,682KB
	model-1000.index	2019-05-26 오전	INDEX 파일	2KB
	model-1000.meta	2019-05-26 오전	META 파일	121KB
	model-1100.data-00000-of-00001	2019-05-26 오전	DATA-00000-OF	73,682KB
	model-1100.index	2019-05-26 오전	INDEX 파일	2KB
	model-1100.meta	2019-05-26 오전	META 파일	121KB
JJ	20102.22.1010100001 000	1200, 1000	J. 25 1000, ac	

Evaluation:

2019-05-26T02:22:50.357046: step 1200, loss 0.37375, acc 0.838

CNN 분류기 평가

eval.py

```
tf.flags.DEFINE_string("imdb_pos_data_file", "./data/test/pos/*", "Data source for the positive data.")
tf.flags.DEFINE_string("imdb_neg_data_file", "./data/test/neg/*", "Data source for the negative data.")
tf.flags.DEFINE_string("checkpoint_dir", "./runs/1522115047/checkpoints", "Checkpoint directory from training run")
tf.flags.DEFINE_boolean("eval_test", True, "Evaluate on all test data")
x_raw, y_test = dh.load_imdb_data_and_labels(FLAGS.imdb_pos_data_file, FLAGS.imdb_neg_data_file)
```

CNN 분류기 평가 (실제 imdb data)

Imdb_eval.py

```
tf.flags.DEFINE_string("real_imdb_x_data_file", "./data/review.txt", "Data source for the positive data.")
tf.flags.DEFINE_string("real_imdb_t__data_file", "./data/score.txt", "Data source for the negative data.")
tf.flags.DEFINE_string("checkpoint_dir", "./runs/1558802980/checkpoints",
"Checkpoint directory from training run")
```

CNN 분류기 평가 (실제 imdb data)

• data_helpers – 함수 추가

```
def load real imdb data and labels(text data file, score data file):
   text_list = list(open(text_data_file, "r", encoding='utf-8').readlines())
   text_list = [s.strip() for s in text_list]
   score_list = list(open(score_data_file, "r", encoding='utf-8').readlines())
   score_list = [s.strip() for s in score_list]
   x_text = [clean_str(sent) for sent in text_list]
   print(score list)
   y = []
   for score in score_list:
      if int(score) > 5:
          y.append([0, 1])
      else:
          y.append([1, 0])
   print(y)
   return [x text, y]
```

CNN 분류기 평가 (실제 imdb data)

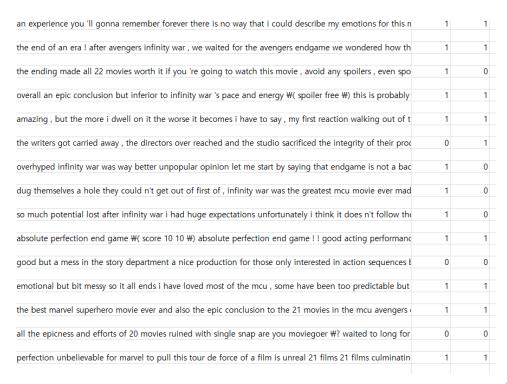
• Imdb_eval.py - 분류 결과 확인

predictions_human_readable = np.column_stack((np.array(x_raw), y_test, all_predictions)) out_path = os.path.join(FLAGS.checkpoint_dir, "..", "prediction_imdb.csv")

Find Movies, TV shows, Celebrities and more.. IMDb Avengers: Endgame (2019) User Reviews Review this title 6.524 Reviews Hide Spoilers Filter by Rating: Show All ▼ Sort by: Helpfulness **10/10** An experience you'll gonna remember forever. Warning: Spoilers 3,229 out of 4,379 found this helpful. Was this review helpful? Sign in to vote Permalink **10/10** The End of an Era! ahmetkozan 25 April 2019 After Avengers Infinity War, we waited for the Avengers Endgame. We wondered how the story would go on, how our heroes would turn back, what would be the end of Thanos. Many theories related to this have been put forward. Avengers Endgame was undoubtedly the most anticipated film of recent years. Normally, the higher the

expectation, the higher the probability of disappointment. But this is not the case for Endgame. Whatever you're expecting, you find much more in the film. This means that

the biggest concern about the film has disappeared.



Q & A