

```
# coding: utf-8
import numpy as np
from common.functions import *
from common.util import im2col, col2im
```

```
class Relu:
```

```
    def __init__(self):
        self.mask = None

    def forward(self, x):
        self.mask = (x <= 0)
        out = x.copy()
        out[self.mask] = 0

        return out

    def backward(self, dout):
        dout[self.mask] = 0
        dx = dout

        return dx
```

```
class Sigmoid:
```

```
    def __init__(self):
        self.out = None

    def forward(self, x):
        out = sigmoid(x)
        self.out = out
        return out

    def backward(self, dout):
        dx = dout * (1.0 - self.out) * self.out

        return dx
```

```
class Affine:
```

```
    def __init__(self, W, b):
        self.W = W
        self.b = b
```

```

self.x = None
self.original_x_shape = None
self.dW = None
self.db = None

def forward(self, x):
    # 张量对应
    self.original_x_shape = x.shape
    x = x.reshape(x.shape[0], -1)
    self.x = x

    out = np.dot(self.x, self.W) + self.b

    return out

def backward(self, dout):
    dx = np.dot(dout, self.W.T)
    self.dW = np.dot(self.x.T, dout)
    self.db = np.sum(dout, axis=0)

    dx = dx.reshape(*self.original_x_shape) # 入力データの形状に戻す(テンソル対応)
    return dx

class SoftmaxWithLoss:
    def __init__(self):
        self.loss = None
        self.y = None # softmax の出力
        self.t = None # 教師データ

    def forward(self, x, t):
        self.t = t
        self.y = softmax(x)
        self.loss = cross_entropy_error(self.y, self.t)

        return self.loss

    def backward(self, dout=1):
        batch_size = self.t.shape[0] # 100
        if self.t.size == self.y.size: # 教師データが one-hot-vector
            dx = (self.y - self.t) / batch_size
        else:
            dx = self.y.copy()
            dx[np.arange(batch_size), self.t] -= 1

```

```
dx = dx / batch_size
```

```
return dx
```

```
class Dropout:
```

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"""
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```
http://arxiv.org/abs/1207.0580
```

```
"""
```

```
def __init__(self, dropout_ratio=0.5):  
    self.dropout_ratio = dropout_ratio  
    self.mask = None
```

```
def forward(self, x, train_flg=True):  
    if train_flg:  
        self.mask = np.random.rand(*x.shape) > self.dropout_ratio  
        return x * self.mask  
    else:  
        return x * (1.0 - self.dropout_ratio)
```

```
def backward(self, dout):  
    return dout * self.mask
```

```
class BatchNormalization:
```

```
"""
```

```
http://arxiv.org/abs/1502.03167
```

```
"""
```

```
def __init__(self, gamma, beta, momentum=0.9, running_mean=None, running_var=None):  
    self.gamma = gamma  
    self.beta = beta  
    self.momentum = momentum  
    self.input_shape = None # Conv 层为四维，全连接层为二维
```

```
# 测试期间使用的平均值和方差  
self.running_mean = running_mean  
self.running_var = running_var
```

```
# backward 時に使用する中間データ  
self.batch_size = None  
self.xc = None  
self.std = None  
self.dgamma = None  
self.dbeta = None
```

```

def forward(self, x, train_flg=True):
    self.input_shape = x.shape
    if x.ndim != 2:
        N, C, H, W = x.shape
        x = x.reshape(N, -1)

    out = self.__forward(x, train_flg)

    return out.reshape(*self.input_shape)

def __forward(self, x, train_flg):
    if self.running_mean is None:
        N, D = x.shape
        self.running_mean = np.zeros(D)
        self.running_var = np.zeros(D)

    if train_flg:
        mu = x.mean(axis=0)
        xc = x - mu
        var = np.mean(xc**2, axis=0)
        std = np.sqrt(var + 10e-7)
        xn = xc / std

        self.batch_size = x.shape[0]
        self.xc = xc
        self.xn = xn
        self.std = std
        self.running_mean = self.momentum * self.running_mean + (1-self.momentum) *
mu
        self.running_var = self.momentum * self.running_var + (1-self.momentum) * var
    else:
        xc = x - self.running_mean
        xn = xc / ((np.sqrt(self.running_var + 10e-7)))

    out = self.gamma * xn + self.beta
    return out

def backward(self, dout):
    if dout.ndim != 2:
        N, C, H, W = dout.shape
        dout = dout.reshape(N, -1)

    dx = self.__backward(dout)

```

```

dx = dx.reshape(*self.input_shape)
return dx

def __backward(self, dout):
    dbeta = dout.sum(axis=0)
    dgamma = np.sum(self.xn * dout, axis=0)
    dxn = self.gamma * dout
    dxc = dxn / self.std
    dstd = -np.sum((dxn * self.xc) / (self.std * self.std), axis=0)
    dvar = 0.5 * dstd / self.std
    dxc += (2.0 / self.batch_size) * self.xc * dvar
    dmu = np.sum(dxc, axis=0)
    dx = dxc - dmu / self.batch_size

    self.dgamma = dgamma
    self.dbeta = dbeta

    return dx

```

class Convolution:

"""

主函数里 Convolution 要赋参数 W 和 b, 这些参数在函数的使用过程中都需要以 self.W, self.b 的形式出现。调用时的格式如 m=Convolution(W,b)[已经赋值的可以不用出现如 stride]

"""

```

def __init__(self, W, b, stride=1, pad=0):
    self.W = W
    self.b = b
    self.stride = stride
    self.pad = pad

    # 中间数据（用于备份）
    self.x = None
    self.col = None
    self.col_W = None

    # 加权，偏置参数的梯度
    self.dW = None
    self.db = None

def forward(self, x):
    FN, C, FH, FW = self.W.shape # 对 FN, C, FH, FW 分别赋值
    N, C, H, W = x.shape # 对 N, C, H, W 分别赋值

```

```
out_h = 1 + int((H + 2*self.pad - FH) / self.stride)
out_w = 1 + int((W + 2*self.pad - FW) / self.stride)
```

```
col = im2col(x, FH, FW, self.stride, self.pad)#输入数据的展开
col_W = self.W.reshape(FN, -1).T # 滤波器的展开
```

```
out = np.dot(col, col_W) + self.b #卷积运算
out = out.reshape(N, out_h, out_w, -1).transpose(0, 3, 1, 2)#变成 N C H W
#输出大小转换为合适的形状
```

```
self.x = x
self.col = col
self.col_W = col_W
```

```
return out
```

```
def backward(self, dout):
    FN, C, FH, FW = self.W.shape
    dout = dout.transpose(0,2,3,1).reshape(-1, FN)

    self.db = np.sum(dout, axis=0)
    self.dW = np.dot(self.col.T, dout)
    self.dW = self.dW.transpose(1, 0).reshape(FN, C, FH, FW)

    dcol = np.dot(dout, self.col_W.T)
    dx = col2im(dcol, self.x.shape, FH, FW, self.stride, self.pad)

    return dx
```

```
class Pooling:
    def __init__(self, pool_h, pool_w, stride=2, pad=0):
        self.pool_h = pool_h
        self.pool_w = pool_w
        self.stride = stride
        self.pad = pad

        self.x = None
        self.arg_max = None

    def forward(self, x):
        N, C, H, W = x.shape
        out_h = int(1 + (H - self.pool_h) / self.stride)
```

```

out_w = int(1 + (W - self.pool_w) / self.stride)

col = im2col(x, self.pool_h, self.pool_w, self.stride, self.pad)
col = col.reshape(-1, self.pool_h*self.pool_w)

arg_max = np.argmax(col, axis=1)
out = np.max(col, axis=1)
out = out.reshape(N, out_h, out_w, C).transpose(0, 3, 1, 2)#原来有多少批池化后不变。

self.x = x
self.arg_max = arg_max

return out

def backward(self, dout):
    dout = dout.transpose(0, 2, 3, 1)

    pool_size = self.pool_h * self.pool_w
    dmax = np.zeros((dout.size, pool_size))
    dmax[np.arange(self.arg_max.size), self.arg_max.flatten()] = dout.flatten()
    dmax = dmax.reshape(dout.shape + (pool_size,))

    dcol = dmax.reshape(dmax.shape[0] * dmax.shape[1] * dmax.shape[2], -1)
    dx = col2im(dcol, self.x.shape, self.pool_h, self.pool_w, self.stride, self.pad)

    return dx

```