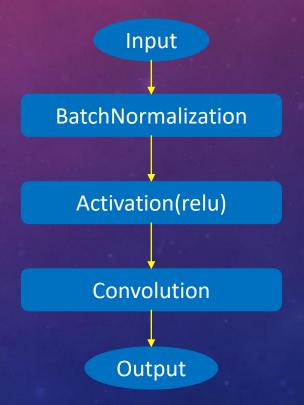


DENSE CONVOLUTIONAL NEURAL NETWORK

- Dense connectivity $X_{l} = H_{l}([X_{0}, X_{1}, X_{2}..., X_{l-1}])$
- Composite Function(transition layer)

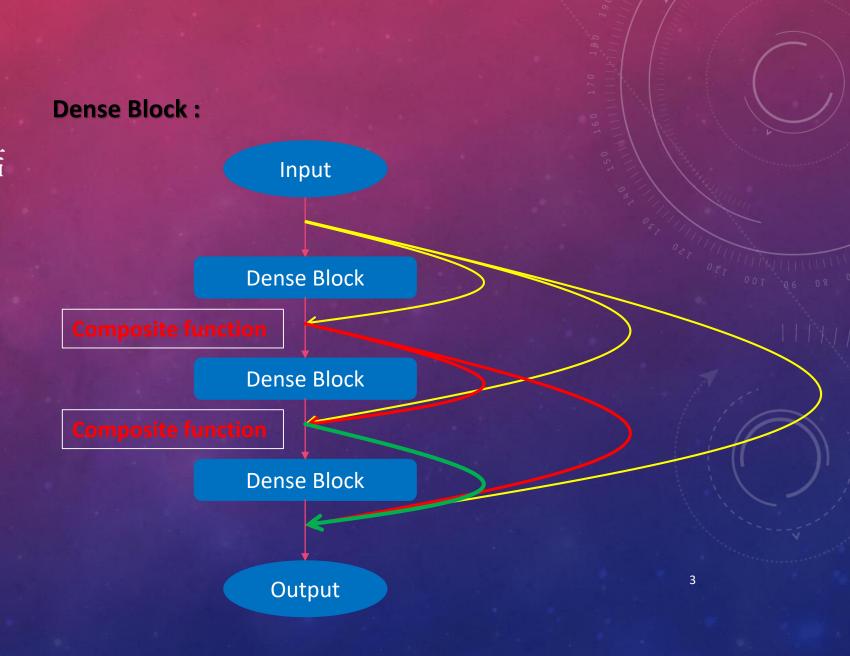
Composite Function:



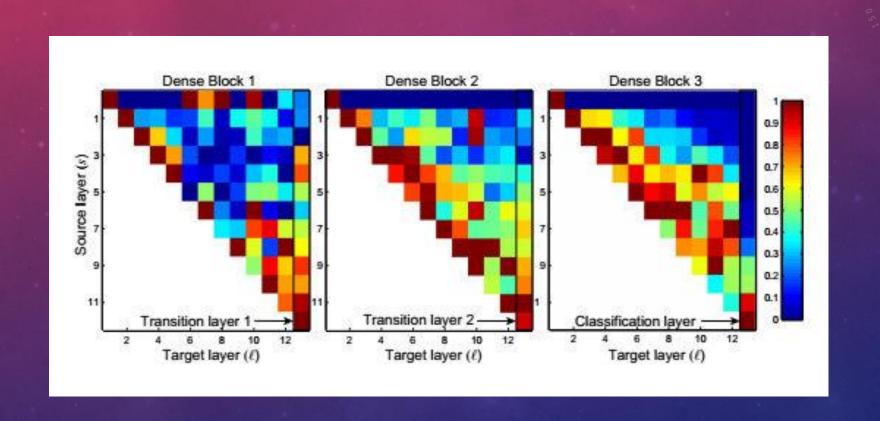
DENSE BLOCK

• L層有L(L+1)/2連結

1+2+3+...+L = L(L+1)/2

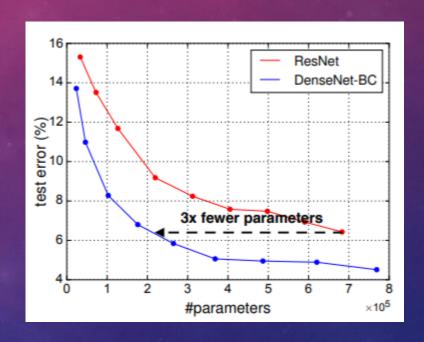


DENSENET 優勢



DENSENET VS RESNET

- Adding vs Concatenation
- ResNet利用identity mapping
- DenseNet使feature更完美利用



截至https://arxiv.org/pdf/1608.06993.pdf



DENSENET-CIFAR10(1/8)

引入所需套件:

```
import keras
from keras.datasets import cifar10
                                                             #本次使用資料集
from keras.models import Model
                                                             #keras建立模型方式
from keras.layers import Dense,Conv2D,Input,AveragePooling2D
                                                             #Dense建立一般層、Conv2D建立CNN層
from keras.layers import Flatten, MaxPooling2D, BatchNormalization #FLatten使CNN過渡到一般層
from keras.layers import Concatenate, Activation
                                                             #利用Concatenate結合feature_map
from keras.optimizers import Adam
                                                             #優什器
from keras.callbacks import ModelCheckpoint,ReduceLROnPlateau
                                                             #可選擇使用,可參考ResNet.py
from keras.utils import to_categorical
                                                             #做one-hot encoding用
```

DENSENET-CIFAR10(2/8)

載入資料集:

```
###---載入資料集---###

(x_train,y_train),(x_test,y_test) = cifar10.load_data()
```

正規化:

```
###---normalization---###

x_train = x_train/255
x_test = x_test/255
```

One-hot encoding:

```
###---one-hot encoding---###

y_train = to_categorical(y_train,10)
y_test = to_categorical(y_test,10)
```

DENSENET-CIFAR10(3/8)

定義Dense Block:

```
###---Dense Block---###

def add_denseblock(inputs,filters):
    x = BatchNormalization()(inputs)
    x = Activation('relu')(x)
    conv = Conv2D(filters,(3,3),padding='same')(x)
    conca = Concatenate(axis=-1)([inputs,conv]) #與ResNet不同之處,ResNet使用add方式將輸入合併
    return conca
```

定義Transition Layer(composite function):

```
###---Transition---###
def add_transition(inputs,filters):
    x = BatchNormalization()(inputs)
    x = Activation('relu')(x)
    x = Conv2D(filters,(1,1),padding='same')(x)
    avg = AveragePooling2D(pool_size=(2,2))(x)
    return avg
```

DENSENET-CIFAR10(4/8)

定義輸出層:

```
###---Output_Layer---###
def output_layer(inputs):
    x = BatchNormalization()(inputs)
    x = Activation('relu')(x)
    avg = AveragePooling2D(pool_size=(2,2))(x)
    y = Flatten()(avg)
    output = Dense(10,activation='softmax')(y)
    return output
```

DENSENET-CIFAR10(5/8)

建立模型:

```
###---建立模型---###
inputs = Input(shape=(32,32,3))
first_conv = Conv2D(32,(3,3),padding='same')(inputs) #第一層CNN
for i in range(3):
                                                    #建立四層dense_block中間有三層transition_layer
   if i == 0:
       block = add_denseblock(first_conv,32)
   else:
       block = add_denseblock(transition,32)
   transition = add_transition(block,32)
block = add_denseblock(transition,32)
                                                    #第四層dense block
output = output_layer(block)
model = Model(inputs=[inputs], outputs=[output])
model.summary()
```

DENSENET-CIFAR10(6/8)

調整參數:

印出結果:

```
###---印出結果---##

score = model.evaluate(x_test,y_test,verbose=1)
print('Test loss:',score[0])
print('Test accuracy:',score[1])

###---Done---###
```



