## Densely Connected Convolutional Networks - Summary

## 1. Abstract-

- The network has L(L+1)/2 direct connections.
- Input of a layer is a feature map of all preceding layers.
- Alleviate the vanishing-gradient problem.
- Encourage feature reuse and substantially reduce the number of parameters.

## 2. Introduction -

- As CNN becomes increasingly deep, the gradient can "wash out" when it reaches the end.
- ResNet, Highway Net, and Fractal Net handle this by "creating short paths from early layers to later layers"
- Dense Net connects all layers directly with each other.
- It combines features by concatenating them. Hence, the lth layer has l inputs, consisting of the feature maps of all preceding convolutional blocks. Its feature maps are passed on to all subsequent L layers. This introduces L(L+1) 2 connections in an L-layer network.
- Densenet Function x = Hl([x0,x1,x2...xl]), [x0,x1,x2...xl] refers to the concatenation of the feature maps produced in layers.
- Hl is a composite function of BN -> ReLU -> 3\*3 Conv 2d.
- The feature maps of previous layers allow it to build on already-learned features rather than learning redundant ones.
- DenseNet doesn't require large feature maps, as information flows more effectively through the network, minimizing parameter usage.
- Growth Rate If each function H produces k feature maps, it follows that the l
  th layer has k0+k\*(l-1) input feature maps, with k = 12 only, k is known as
  growth rate.
- Each layer has direct access to the loss function and original input.
- Bottle Neck Layers Introduces 1\*1 conv before every 3\*3 conv, which reduces the number of features.
  - BN-ReLU-Conv(1 1)-BN-ReLU-Conv(3 3), known as DenseNet B.
- Compression Factor

  If a dense block contains m feature maps, we let the following transition layer

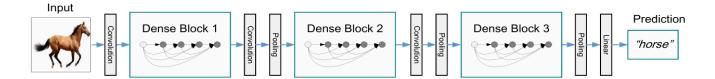
generate m output feature maps, where 0< Theta <=1 is referred to as the compression factor.

We refer the DenseNet with Theta<1 as DenseNet-C, and we set Theta=0.5 in our experiment.

Dense Net BC uses both Bottle Neck layers and Compression Factor.

## 3. Implementation Details

- It has 3 dense blocks with equal number of layers
- Before entering the dense block a convolution of 16 output channels is implemented.
- Transition Layer 1\*1 conv -> 2\*2 averagepool 2d
- At the end, the Gloabal Avg pool is followed by the soft max.



Output Size	DenseNet-121	DenseNet-169	DenseNet-201	DenseNet-264
112 × 112	$7 \times 7$ conv, stride 2			
56 × 56	$3 \times 3$ max pool, stride 2			
nse Block (1) $56 \times 56$	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 6}$	[1×1 conv]	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 6}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \\ \times 6 \end{bmatrix}$
	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 0}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 0}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 6}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 6}$
56 × 56	1 × 1 conv			
$28 \times 28$	$2 \times 2$ average pool, stride 2			
28 × 28	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 12}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 12}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 12}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \\ \times 12 \end{bmatrix}$
	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{-12}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{-12}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{-12}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{12}$
$28 \times 28$	1 × 1 conv			
14 × 14	$2 \times 2$ average pool, stride 2			
14 × 14	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 24}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 32}$	[ 1 × 1 conv ] × 48	$\begin{bmatrix} 1 \times 1 \text{ conv} \\ \times 64 \end{bmatrix}$
14 / 14	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{24}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{32}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{46}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}$
14 × 14	1 × 1 conv			
7 × 7	$2 \times 2$ average pool, stride 2			
7 . 7	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 16}$	$\begin{bmatrix} 1 \times 1 \text{ conv} \end{bmatrix}_{\times 32}$	[ 1 × 1 conv ] × 32	$\begin{bmatrix} 1 \times 1 \text{ conv} \\ 1 \times 48 \end{bmatrix}$
' ^ '	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 10}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 32}$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix} \times 32$	$\begin{bmatrix} 3 \times 3 \text{ conv} \end{bmatrix}^{\times 46}$
1 × 1	$7 \times 7$ global average pool			
	1000D fully-connected, softmax			
	$     \begin{array}{r}       112 \times 112 \\       56 \times 56 \\       \hline       56 \times 56 \\       \hline       56 \times 56 \\       \hline       28 \times 28 \\       \hline       28 \times 28 \\       \hline       14 \times 14 \\       \hline       14 \times 14 \\       \hline       7 \times 7 \\       \hline       7 \times 7     \end{array} $	$     \begin{array}{c cccc}         & 112 \times 112 \\         & 56 \times 56 \\         & 56 \times 56 \\         & 28 \times 28 \\         & 28 \times 28 \\         & 28 \times 28 \\         & 14 \times 14 \\         & 14 \times 14 \\         & 7 \times 7 \\         & 7 \times 7 \\         &   &   &   &   &   &   &   &$	$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$