CNN Architecture Summary

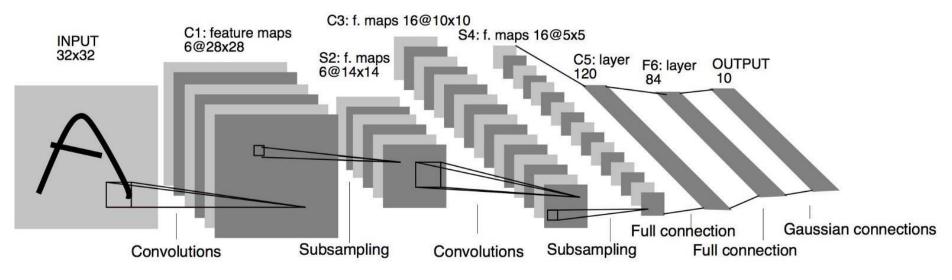
NCSOFT Vision AI Lab

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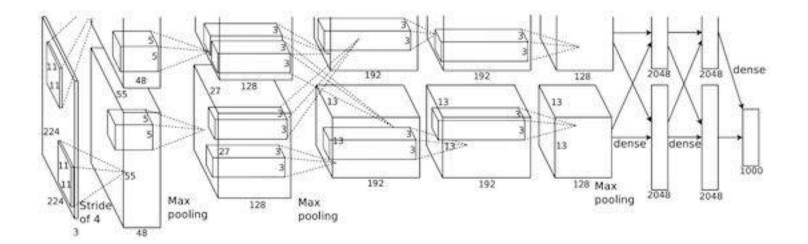


LeNet5



| Characteristics | Contribution | Etc |
|--|---|------------------------|
| Conv – Pooling – Activation | Use of convolution to extra ct spatial features | Slow to train |
| Average pooling | · | Hard to train (Neurons |
| Sigmoid activation & tanh activation | Subsample using spatial average of maps | dies quickly) |
| • 5x5 Convolution filter | | |
| 7 layers and less than IM parameters | | |

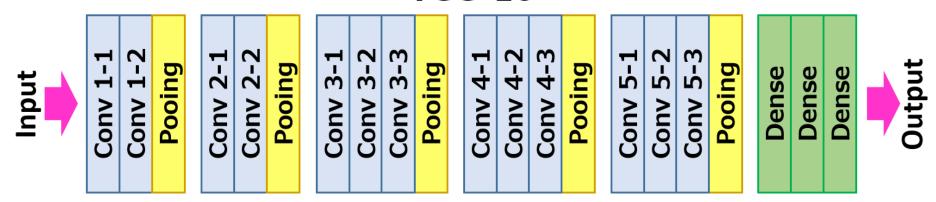
AlexNet



| Characteristics | Contribution | Etc |
|-----------------------------------|------------------------------|-----|
| • IIxII, 5x5, and 3x3 convolution | Relu activation | |
| Max pooling | Dropout for regularization | |
| • 3 FC | Local response normalization | |
| • 60M parameters | Image augmentation | |
| | Multiple GPU | |

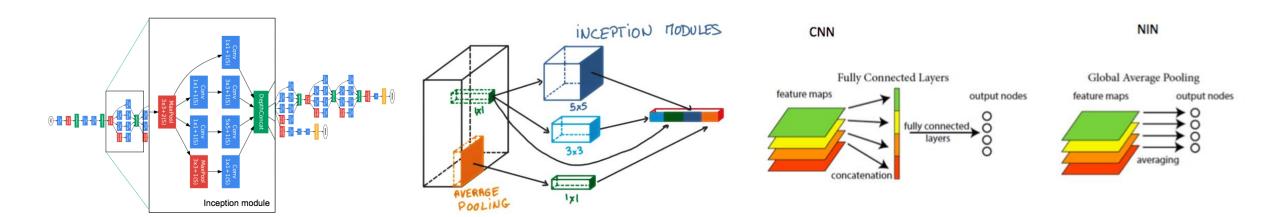
VggNet

VGG-16



| Characteristics | Contribution | Etc |
|-------------------|---------------------|-----------------|
| • 3x3 conv | Simple architecture | Many parameters |
| Max pooling | • Easy to use | |
| • I30M parameters | Practical | |
| | | |
| | | |

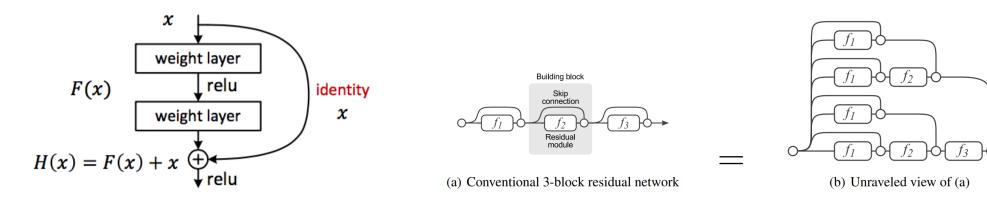
GoogLeNet



| Characteristics | Contribution | Etc |
|------------------------|---------------------------|--------------------------|
| • x convolution | Depth reduction using IxI | Hard to use |
| Global average pooling | | • Inception v2, v3, v4 ~ |
| • 6.8 M parameters | | |
| | | |
| | | |

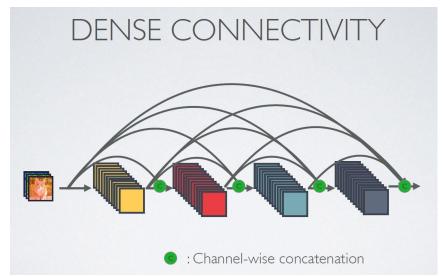
ResNet

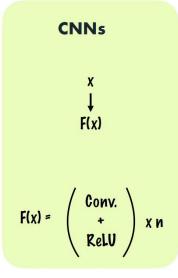
Residual net

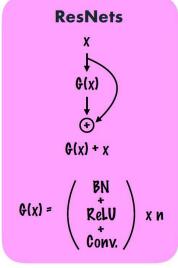


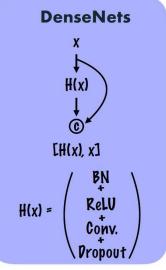
| Characteristics | Contribution | Etc |
|-----------------|-----------------|-----------------------|
| Going deep | Skip connection | Pre-activation resnet |
| • IM parameters | • Easy to use | • ResNeXt |
| | Practical | |
| | | |
| | | |

DenseNet



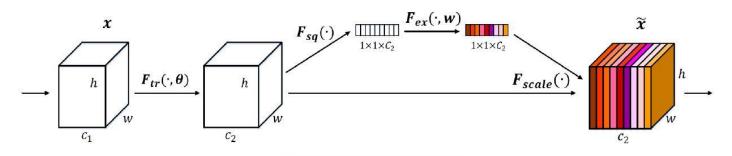






| Characteristics | Contribution | Etc |
|--------------------|-----------------|-------------|
| • BN – Relu – conv | • Concatenation | More memory |
| • IM parameters | • Easy to use | |
| | Practical | |
| | | |
| | | |

SENet



Squeeze

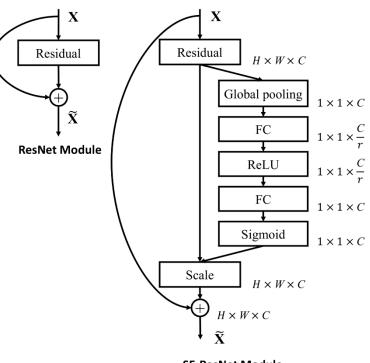
- Shrinking feature maps $\in \mathbb{R}^{w \times h \times c_2}$ through spatial dimensions $(w \times h)$
- Global distribution of channelwise responses

Excitation

- Learning $W \in \mathbb{R}^{c_2 \times c_2}$ to explicitly model channel-association
- Gating mechanism to produce channel-wise weights

Scale

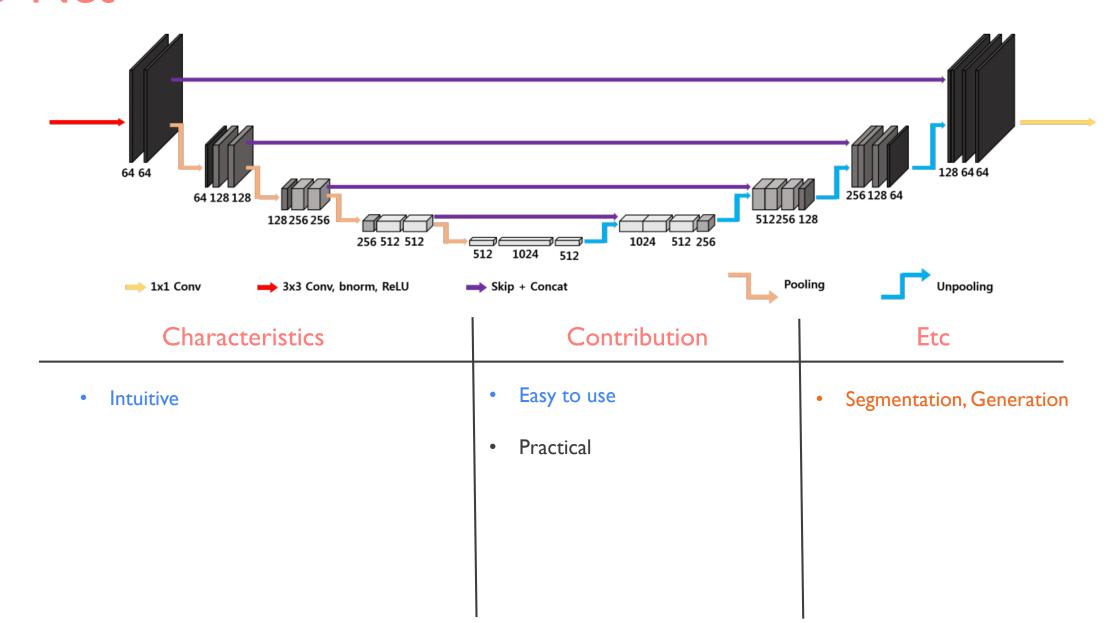
ullet Reweighting the feature maps $\in \mathbb{R}^{w imes h imes c_2}$



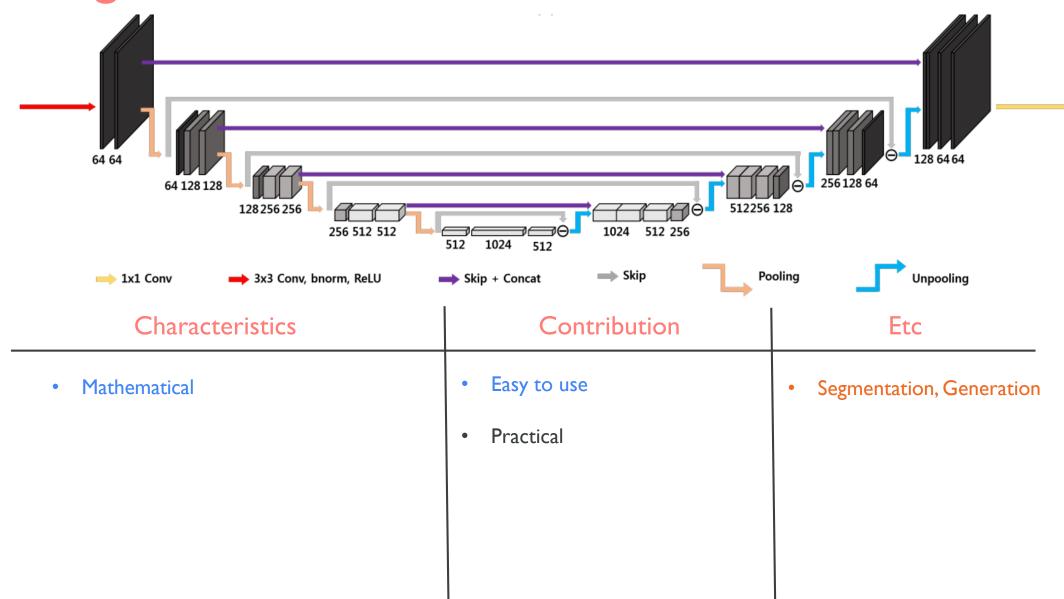
SE-ResNet Module

| Characteristics | Contribution | Etc |
|-----------------|---------------|-----|
| • Attention | • Easy to use | |
| | Practical | |
| | | |
| | | |
| | | |

U-Net



Framing U-Net



END

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김준호