# Resnet Architecture

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## Basic Block

• 18/34 layer

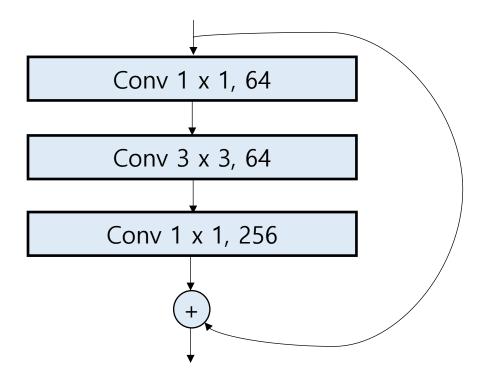
```
Conv 3 x 3, 64

Conv 3 x 3, 256
```

```
class BasicBlock(nn.Module):
    expansion = 1
    def init (self, inplanes, planes, stride=1, downsample=None):
        super(BasicBlock, self). init ()
        self.conv1 = conv3x3(inplanes, planes, stride)
        self.bn1 = nn.BatchNorm2d(planes)
        self.relu = nn.ReLU(inplace=True)
        self.conv2 = conv3x3(planes, planes)
        self.bn2 = nn.BatchNorm2d(planes)
        self.downsample = downsample
        self.stride = stride
    def forward(self, x):
        identity = x
        #x.shape= 3,64,64라고 가정
        out = self.conv1(x) # 3x3 stride = 2
        #outdl 3,32,32가 될 것임.
        out = self.bnl(out)
        out = self.relu(out)
        out = self.conv2(out) # 3x3 stride = 1
        out = self.bn2(out)
        if self.downsample is not None: #downsample은 stride가 2가 들어오는 경우
           identity = self.downsample(x)
        out += identity
        out = self.relu(out)
        return out
```

## Bottleneck Block

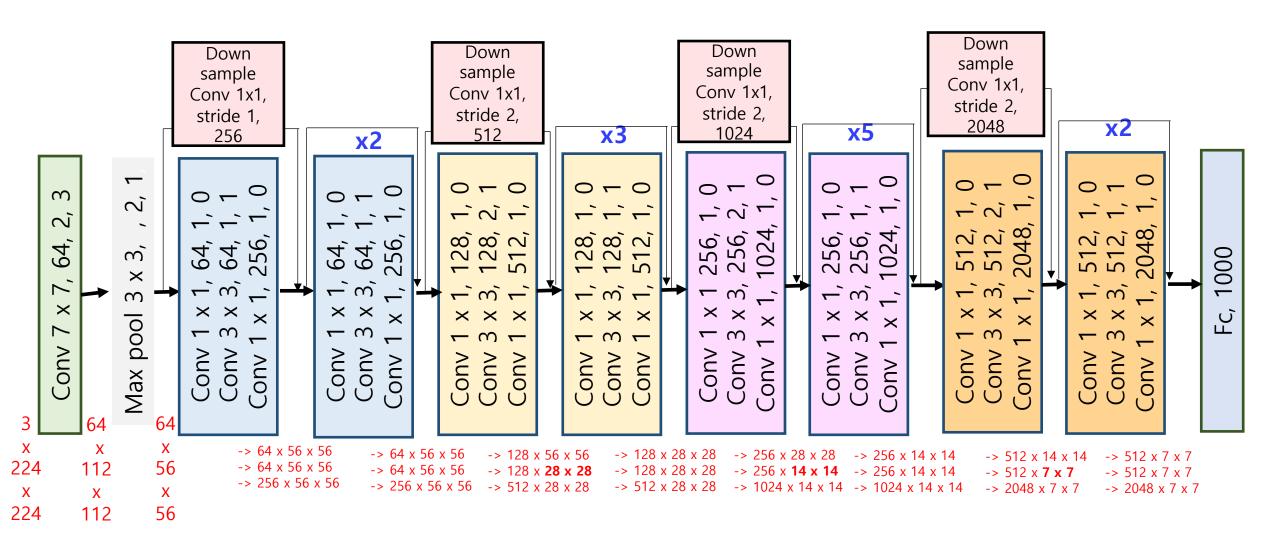
50/101/153 layer에 대해서는 bottleneck 구조를 사용하여 conv 1x1로 차원을 줄였다가 늘려서 연산량을 절감시킴



```
class Bottleneck(nn.Module):
    expansion = 4
    def init (self, inplanes, planes, stride=1, downsample=None):
        super(Bottleneck, self). init ()
        self.conv1 = conv1x1(inplanes, planes) #conv1x1(64,64)
        self.bn1 = nn.BatchNorm2d(planes)
        self.conv2 = conv3x3(planes, planes, stride)#conv3x3(64,64)
        self.bn2 = nn.BatchNorm2d(planes)
        self.conv3 = convlx1(planes, planes * self.expansion) #convlx1(64,256)
        self.bn3 = nn.BatchNorm2d(planes * self.expansion)
        self.relu = nn.ReLU(inplace=True)
        self.downsample = downsample
        self.stride = stride
    def forward(self, x):
        identity = x
        out = self.conv1(x) # 1x1 stride = 1
        out = self.bnl(out)
        out = self.relu(out)
        out = self.conv2(out) # 3x3 stride = stride
        out = self.bn2(out)
        out = self.relu(out)
        out = self.conv3(out) # 1x1 stride = 1
        out = self.bn3(out)
        if self.downsample is not None:
            identity = self.downsample(x)
        out += identity
        out = self.relu(out)
        return out
```

## Resnet 50 architecture

Kernel, out\_size, kernel, stride, padding  $(conv1) + 3 \times (3 + 4 + 6 + 3) + (fc) = 48 + 2 = 50$ 



#### Resnet architecture

Kernel, out\_size, kernel, stride, padding

```
self.inplanes = 64
                   #input: 3*224*224
    \infty 
          ζ
                    self.conv1 = nn.Conv2d(3, 64, kernel size=7, stride=2, padding=3, bias=False)
                   # output: 64*112*112
   64,
                    self.bn1 = nn.BatchNorm2d(64)
          \infty
                   self.relu = nn.ReLU(inplace=True)
          \times
          \mathcal{C}
                   #input: [64,112,112]
                    self.maxpool = nn.MaxPool2d(kernel_size=3, stride=2, padding=1)
   Onv
                   #output: [64,56,56]
          Max
             64
224
             56
     112
224
     112
             56
```

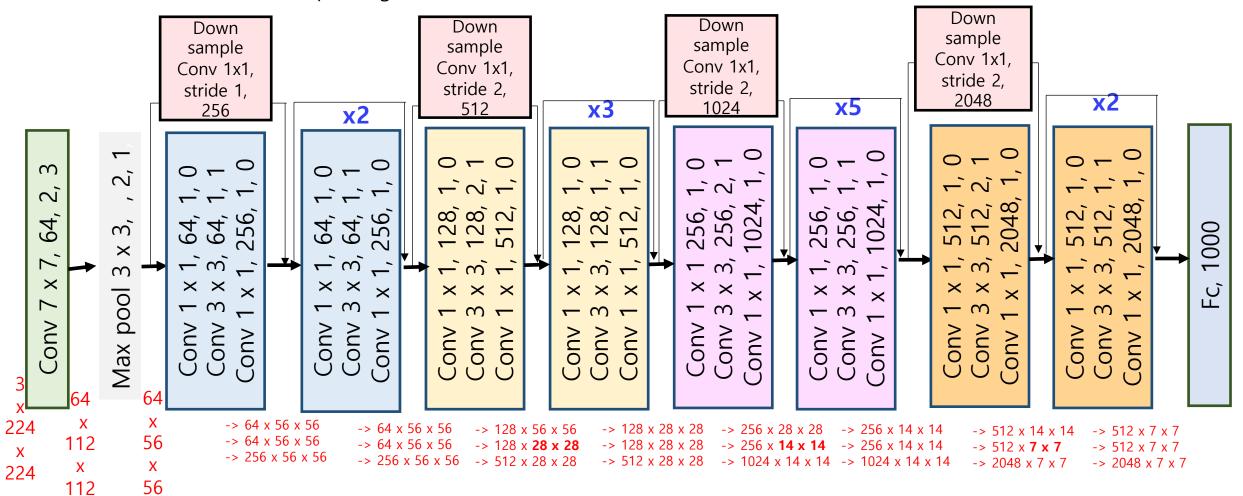
#### Resnet architecture

```
self.layer1 = self._make_layer(bl1 64, layers[0]) #3
```

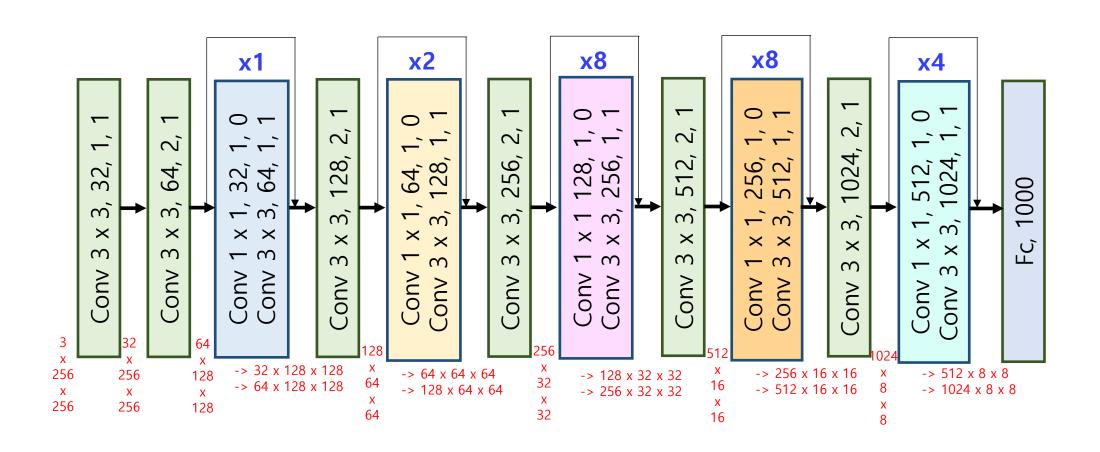
```
# self.layer1 = self. make_layer(Bottleneck, 64, layers[0], 3)
    Kernel, out size, kernel, stride, padding
                                                            def make layer(self, block, planes, blocks, stride=1):
                          Down
                                                                downsample = None
                         sample
                                                                if stride != 1 or self.inplanes != planes * block.expansion: # 64*4(이 코드에서는 채널 수 맞추기 위해서도 사용)
                        Conv 1x1,
                                                                    print(self.inplanes)
                        stride 1,
                                                                    downsample = nn.Sequential(
                           256
                                              x2
                                                                         convlx1(self.inplanes, planes * block.expansion, stride), #convlx1(64, 64*4, 1) #stride=1
                                                                         nn.BatchNorm2d(planes * block.expansion), #batchnrom2d(256)
     \infty 
              2,
    7
                                                                layers = []
                                                                layers.append(block(self.inplanes, planes, stride, downsample))
    64,
                               99
                                          64,
64,
256,
                        64,
64,
               \infty
               ×
                                                                self.inplanes = planes * block.expansion #self.inplanes = 64 * 4 = 256
                        ر س <del>ر</del>
                                           T, W, T,
               \mathcal{C}
     ×
                        \times \times
                                                                 for in range(1, blocks):
                               ×
               Max pool
                                          - ∞ -
                       − ∞ −
                                                                    layers.append(block(self.inplanes, planes)) # * 2
    Conv
                        Conv
                           Conv
                                           Conv
                                              Conv
                                                                return nn.Sequential(*layers)
                                                                #Bottleneck(64,64,1, downsample) / Bottlneck(256,64) / Bottlneck(256,64)
                                                                #Bottleneck(256,128,2, downsample) / Bottlneck(512,128) / Bottlneck(512,128) / Bottlneck(512,128)
                              -> 64 x 56 x 56
224
                                                 -> 64 x 56 x 56
        112
                   56
                              -> 64 x 56 x 56
                                                 -> 64 x 56 x 56
                              -> 256 x 56 x 56
                                                 -> 256 x 56 x 56
224
        112
                   56
```

### Resnet 50 architecture

Kernel, out\_size, kernel, stride, padding



## YOLOv3 Architecture



## YOLOv3 Architecture

