



CS 329P: Practical Machine Learning (2021 Fall)

Residual Connections

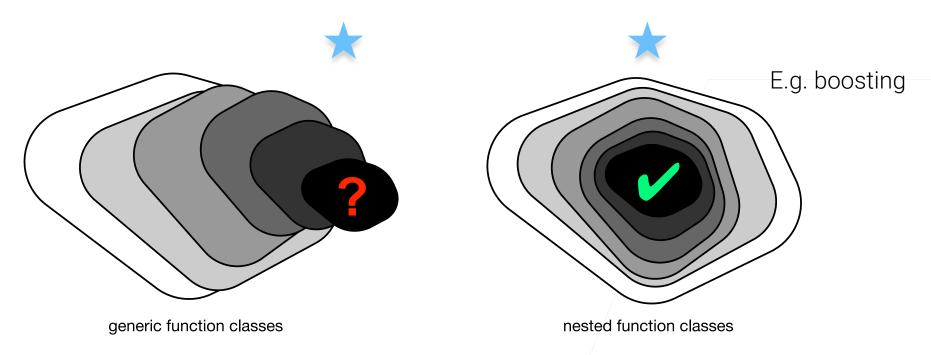
Qingqing Huang, Mu Li, Alex Smola

https://c.d2l.ai/stanford-cs329p

Does Adding Layers Improve Accuracy?



Add more layers may decrease accuracy



Residual Connections



- Adding a layer f changes the function class to f(g(x))
- We want to add to the function class f(g(x)) + g(x)
 - Larger gradient, easier to train

$$\frac{\partial f(g(x))}{\partial x} = \frac{\partial f(g(x))}{\partial g(x)} \frac{\partial g(x)}{\partial x} \qquad \frac{\partial (f(g(x)) + g(x))}{\partial x} = \frac{\partial f(g(x))}{\partial x} + \frac{\partial g(x)}{\partial x}$$

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Boosting over features vs boosting over labels discussed before

ResNet - de facto CNN Architecture

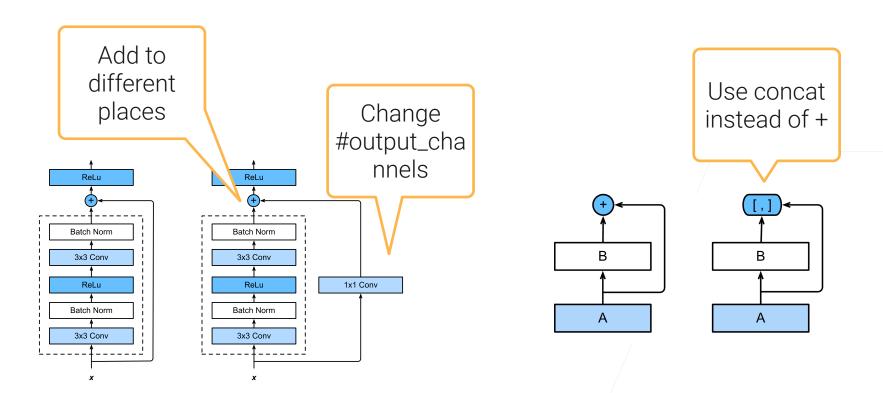


CNN with stacked Residual blocks

```
class ResidualBlock(nn.Module):
    def __init__(self, input_channels, num channels):
        super().__init__()
        self.conv1 = nn.Conv2d(
          input_channels, num_channels, kernel_size=3, padding=1)
        self.conv2 = nn.Conv2d(
          num_channels, num_channels, kernel_size=3, padding=1)
        self.bn1 = nn.BatchNorm2d(num channels)
        self.bn2 = nn.BatchNorm2d(num_channels)
    def forward(self, X):
                                                              Full code: <a href="http://d2l.ai/">http://d2l.ai/</a>
        Y = F.relu(self.bn1(self.conv1(X)))
                                                              chapter_convolutional-modern/
        Y = self.bn2(self.conv2(Y))
        return F.relu(Y + X)
                                                              resnet.html
```

Variants





Summary



- Residual connection allows to pass internal layers
 - Make very deep NN possible (peopled tried CNNs with 1K layers)
 - Build nested function classes
 - Boosting over the features