ML/DL for Everyone with PYTORCH

Lecture 10: CNN



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Other slides: http://bit.ly/PyTorchZeroAll

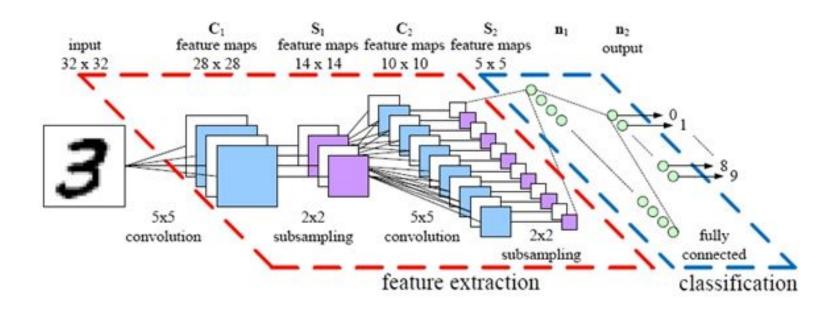


ML/DL for Everyone with PYTORCH

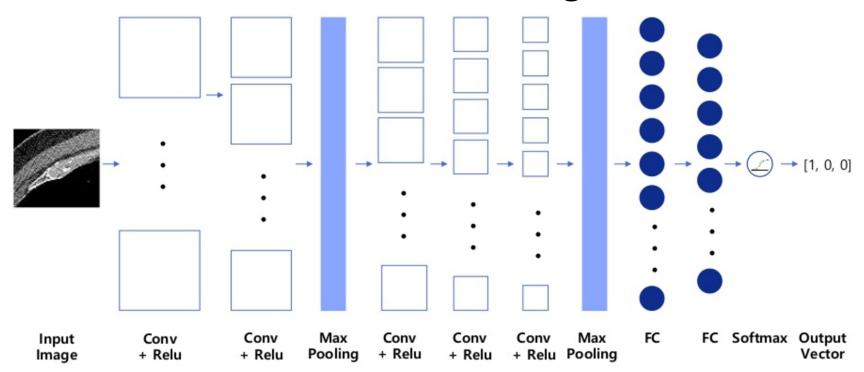
Lecture 10: CNN



CNN

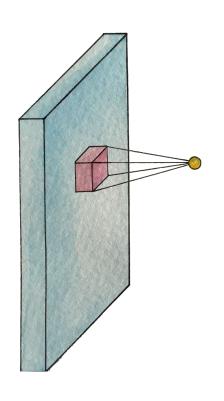


CNN for CT images



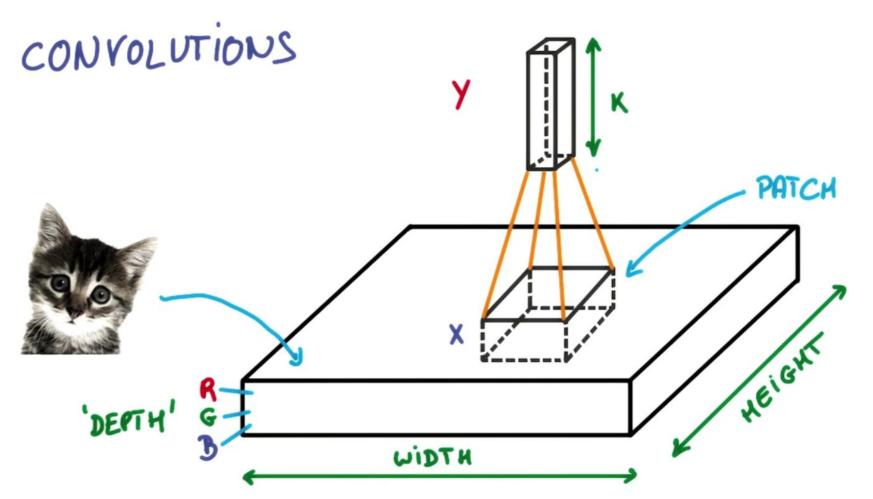
Asan Medical Center & Microsoft Medical Bigdata Contest Winner by GeunYoung Lee and Alex Kim https://www.slideshare.net/GYLee3/ss-72966495

Convolution layer and max pooling



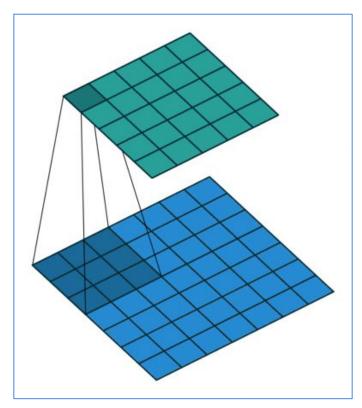
Single depth slice

1	1	2	4
5	6	7	8
3	2	1	0
1	2	3	4

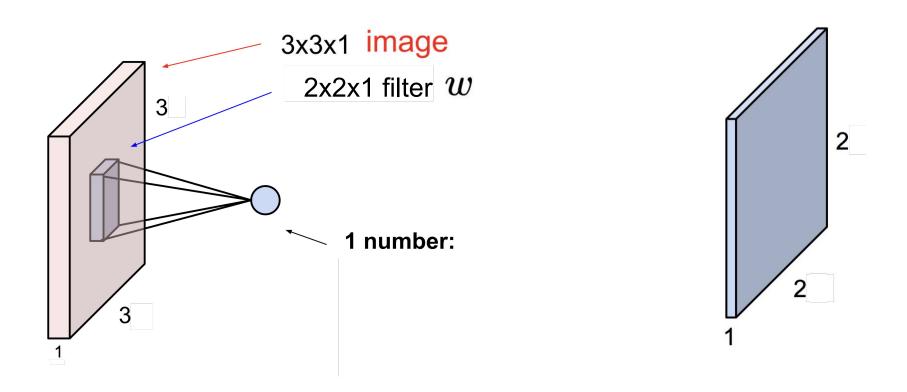


https://ireneli.eu/2016/02/03/deep-learning-05-talk-about-convolutional-neural-network%EF%BC%88cnn%EF%BC%89/

Convolution in Action

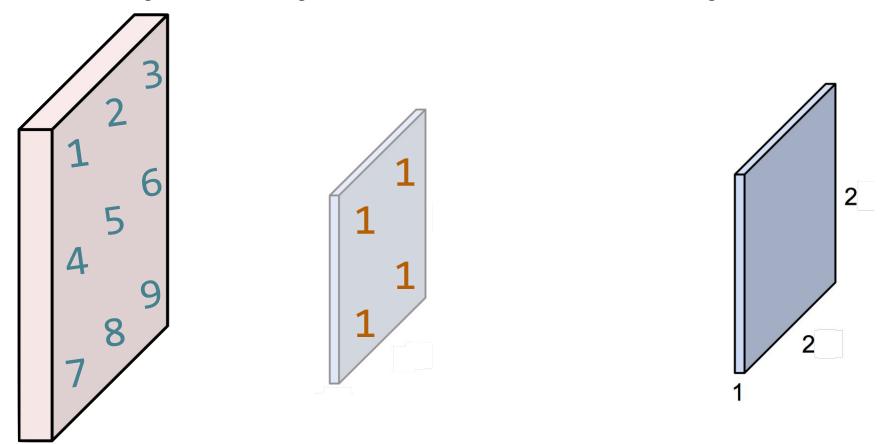


Simple convolution layer Stride: 1x1



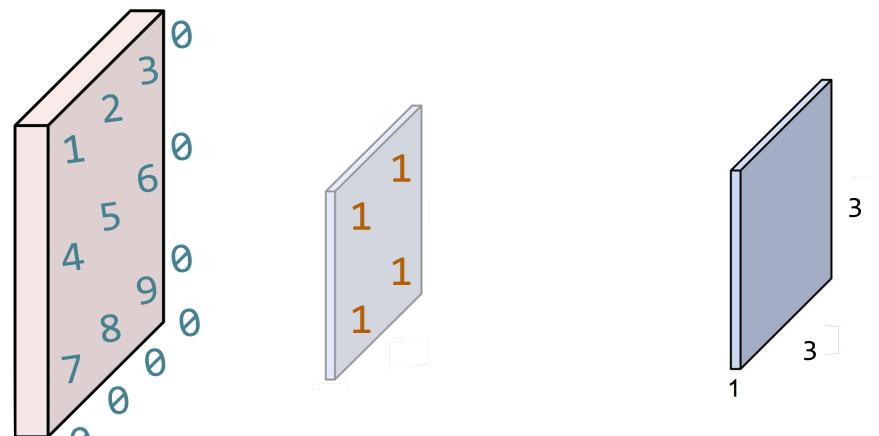
Simple convolution layer

Image: 1,3,3,1 image, Filter: 2,2,1,1, Stride: 1x1, Padding: VALID

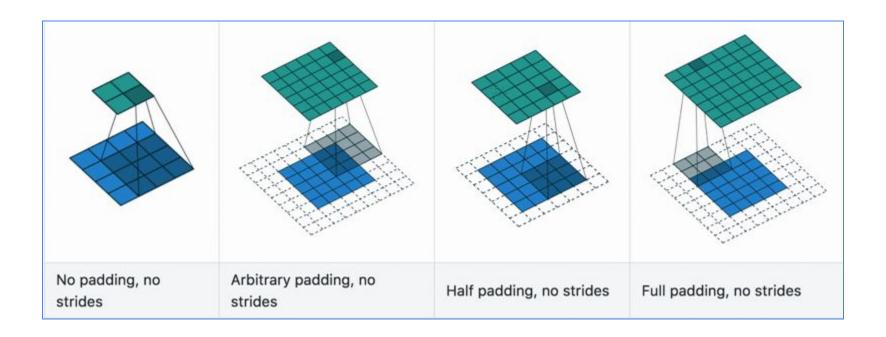


Simple convolution layer

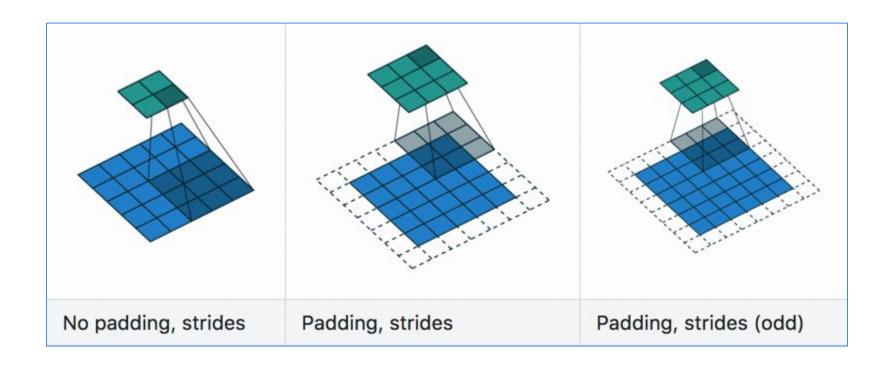
Image: 1,3,3,1 image, Filter: 2,2,1,1, Stride: 1x1, Padding: SAME



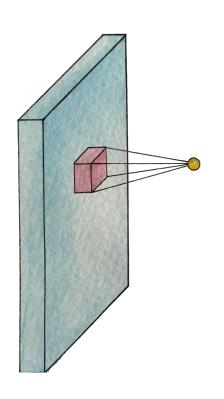
Convolution with padding in Action



Convolution with stride in Action



Max pooling



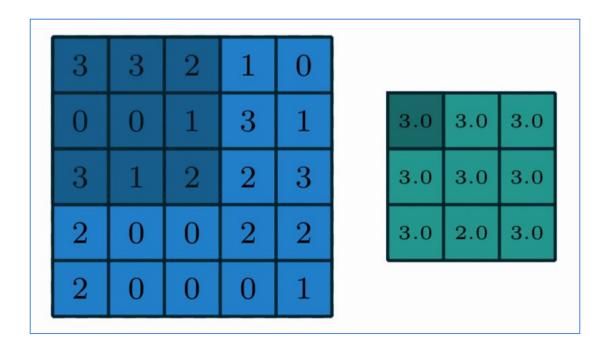
Single depth slice

1	1	2	4
5	6	7	8
3	2	1	0
1	2	3	4

max pool with 2x2 filters and stride 2

6	8		
3	4		

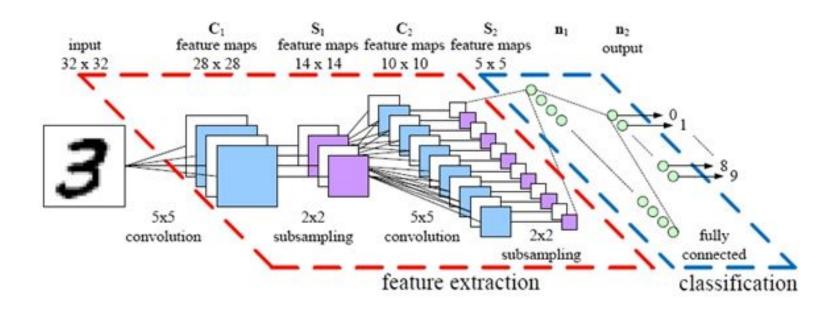
Max Pooling in Action



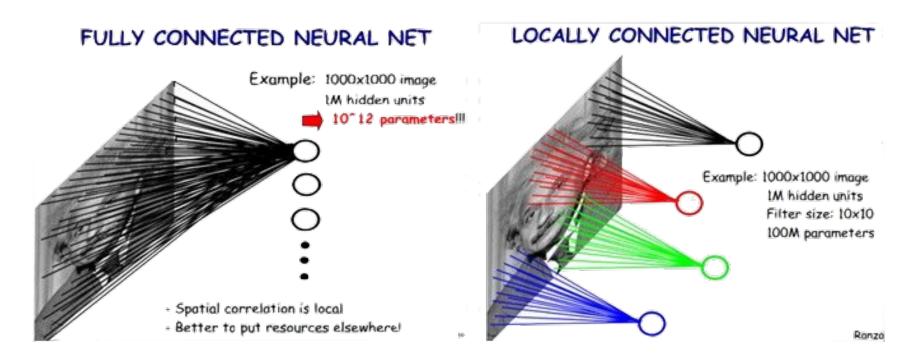
Avg Pooling in Action

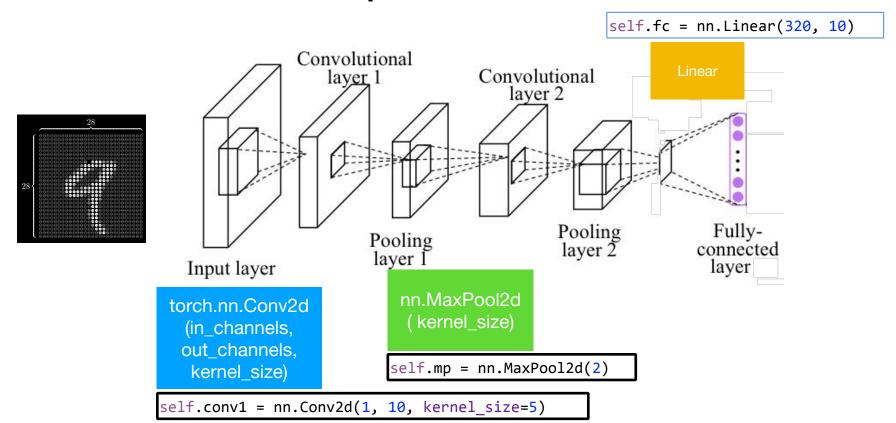
3	3	2	1	0			
0	0	1	3	1	1.7	1.7	1.7
3	1	2	2	3	1.0	1.2	1.8
2	0	0	2	2	1.1	0.8	1.3
2	0	0	0	1			

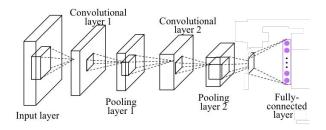
CNN



Locally Connected Features

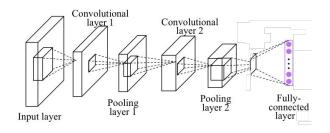








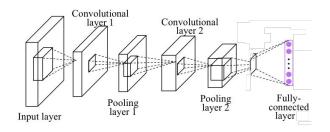
```
class Net(nn.Module):
   def init (self):
       super(Net, self).__init__()
       self.conv1 = nn.Conv2d(1, 10, kernel size=5)
       self.conv2 = nn.Conv2d(10, 20, kernel size=5)
       self.mp = nn.MaxPool2d(2)
       self.fc = nn.Linear(100???, 10) # ??? -> 10
   def forward(self, x):
       in size = x.size(0)
       x = F.relu(self.mp(self.conv1(x)))
       x = F.relu(self.mp(self.conv2(x)))
       x = x.view(in size, -1) # flatten the tensor
       x = self.fc(x)
       return F.log softmax(x)
```





```
class Net(nn.Module):
   def init (self):
       super(Net, self). init ()
       self.conv1 = nn.Conv2d(1, 10, kernel size=5)
       self.conv2 = nn.Conv2d(10, 20, kernel size=5)
       self.mp = nn.MaxPool2d(2)
       self.fc = nn.Linear(100???, 10) # ??? -> 10
   def forward(self, x):
       in size = x.size(0)
       x = F.relu(self.mp(self.conv1(x)))
       x = F.relu(self.mp(self.conv2(x)))
       x = x.view(in_size, -1) # flatten the tensor
       x = self.fc(x)
       return F.log softmax(x)
```

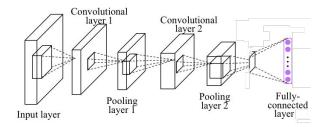
RuntimeError: size mismatch, m1: [64 x **320**], m2: [100 x 10]





```
class Net(nn.Module):
    def init (self):
        super(Net, self). init ()
        self.conv1 = nn.Conv2d(1, 10, kernel size=5)
        self.conv2 = nn.Conv2d(10, 20, kernel size=5)
        self.mp = nn.MaxPool2d(2)
        self.fc = nn.Linear(\frac{320}{10}, 10) # 320 -> 10
    def forward(self, x):
        in size = x.size(0)
        x = F.relu(self.mp(self.conv1(x)))
        x = F.relu(self.mp(self.conv2(x)))
        x = x.view(in_size, -1) # flatten the tensor
        x = self.fc(x)
        return F.log softmax(x)
```

RuntimeError: size mismatch, m1: [64 x **320**], m2: [100 x 10]

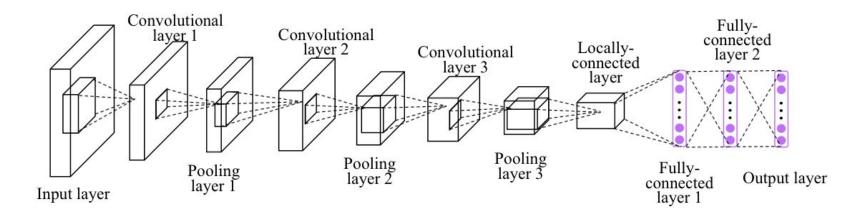




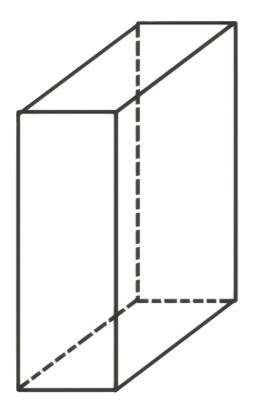
```
class Net(nn.Module):
   def init (self):
       super(Net, self). init ()
       self.conv1 = nn.Conv2d(1, 10, kernel size=5)
       self.conv2 = nn.Conv2d(10, 20, kernel size=5)
       self.mp = nn.MaxPool2d(2)
       self.fc = nn.Linear(320, 10) # 320 -> 10
   def forward(self, x):
       in size = x.size(0)
       x = F.relu(self.mp(self.conv1(x)))
       x = F.relu(self.mp(self.conv2(x)))
       x = x.view(in size, -1) # flatten the tensor
       x = self.fc(x)
       return F.log softmax(x)
```

```
Train Epoch: 9 [46080/60000 (77%)]
                                         Loss: 0.108415
Train Epoch: 9 [46720/60000 (78%)]
                                         Loss: 0.140700
Train Epoch: 9 [47360/60000 (79%)]
                                         Loss: 0.090830
Train Epoch: 9 [48000/60000 (80%)]
                                         Loss: 0.031640
Train Epoch: 9 [48640/60000 (81%)]
                                         Loss: 0.014934
Train Epoch: 9 [49280/60000 (82%)]
                                         Loss: 0.090210
Train Epoch: 9 [49920/60000 (83%)]
                                         Loss: 0.074975
Train Epoch: 9 [50560/60000 (84%)]
                                         Loss: 0.058671
Train Epoch: 9 [51200/60000 (85%)]
                                         Loss: 0.023464
Train Epoch: 9 [51840/60000 (86%)]
                                         Loss: 0.018025
Train Epoch: 9 [52480/60000 (87%)]
                                         Loss: 0.098865
Train Epoch: 9 [53120/60000 (88%)]
                                         Loss: 0.013985
Train Epoch: 9 [53760/60000 (90%)]
                                         Loss: 0.070476
Train Epoch: 9 [54400/60000 (91%)]
                                         Loss: 0.065411
Train Epoch: 9 [55040/60000 (92%)]
                                         Loss: 0.028783
Train Epoch: 9 [55680/60000 (93%)]
                                         Loss: 0.008333
Train Epoch: 9 [56320/60000 (94%)]
                                         Loss: 0.020412
Train Epoch: 9 [56960/60000 (95%)]
                                         Loss: 0.036749
Train Epoch: 9 [57600/60000 (96%)]
                                         Loss: 0.163087
Train Epoch: 9 [58240/60000 (97%)]
                                         Loss: 0.117539
Train Epoch: 9 [58880/60000 (98%)]
                                         Loss: 0.032256
Train Epoch: 9 [59520/60000 (99%)]
                                         Loss: 0.026360
Test set: Average loss: 0.0483, Accuracy: 9846/10000 (98%)
```

Exercise 10-1: Implement CNN more layers



INCEPTION MODULES



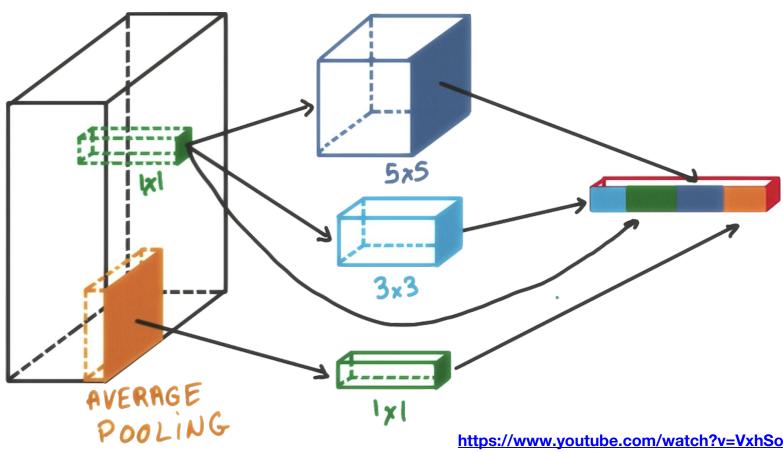


INCEPTION MODULES 3x3 1

POOLING?

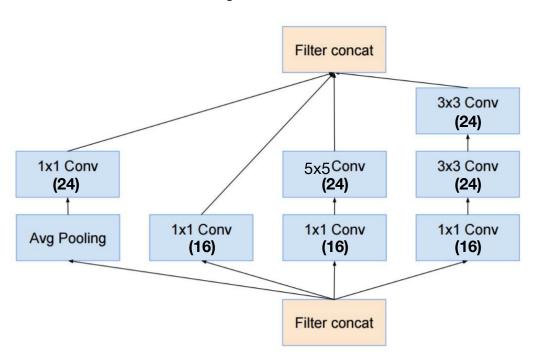
5x5 ?

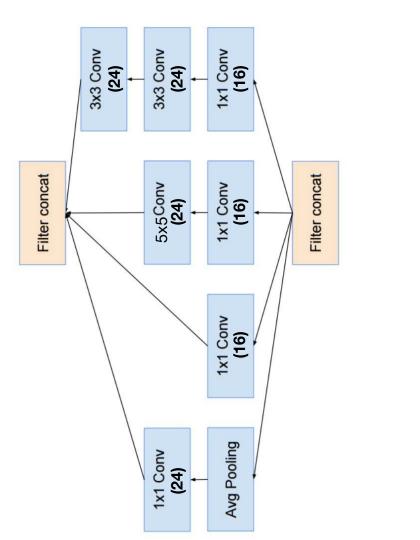
INCEPTION MODULES

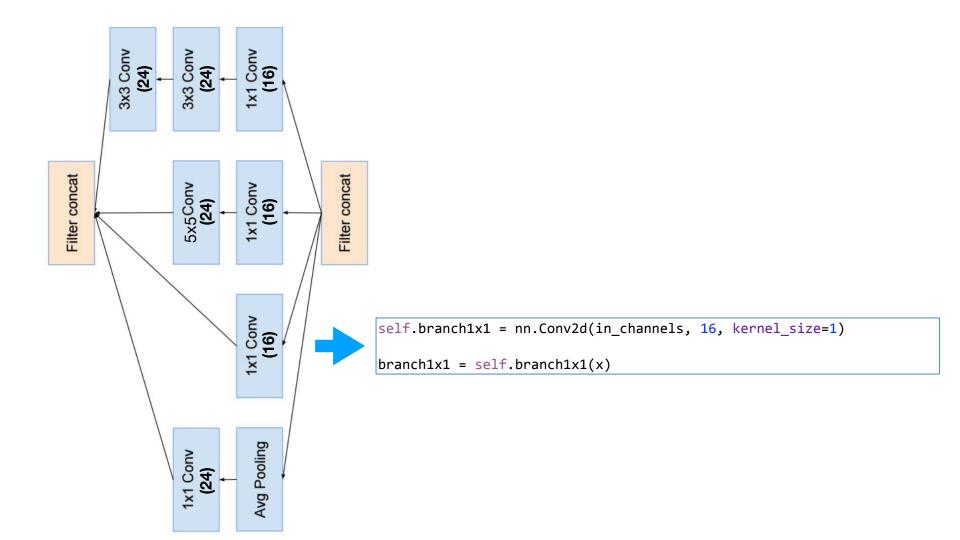


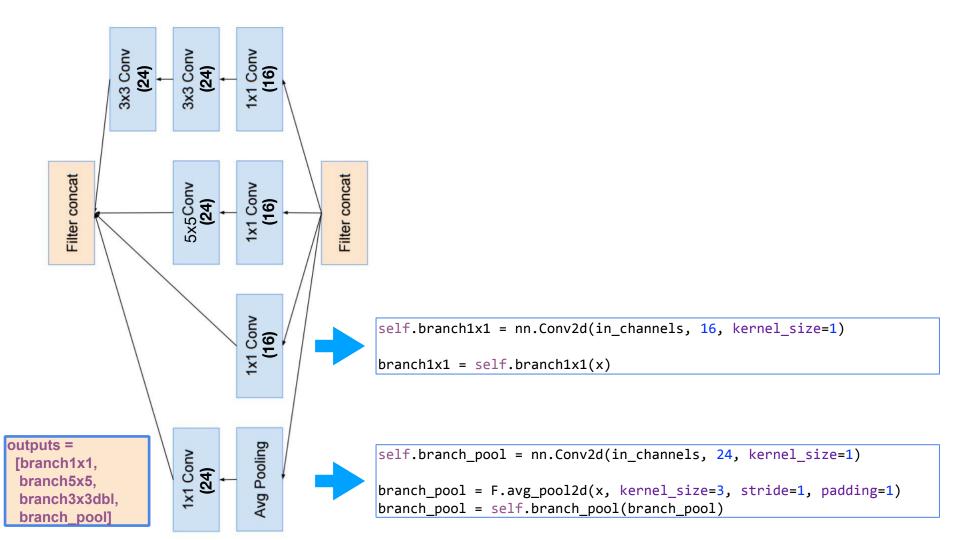
https://www.youtube.com/watch?v=VxhSouuSZDY

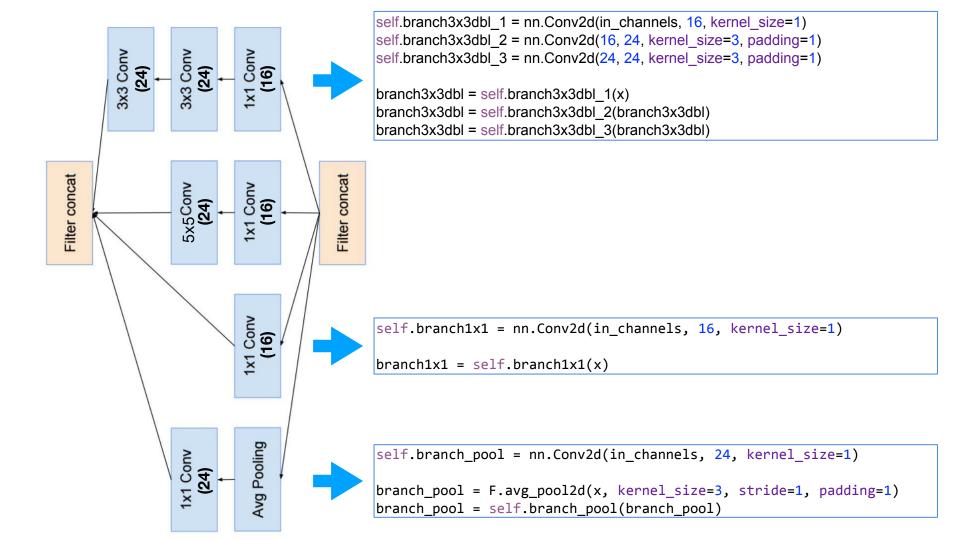
Inception Module

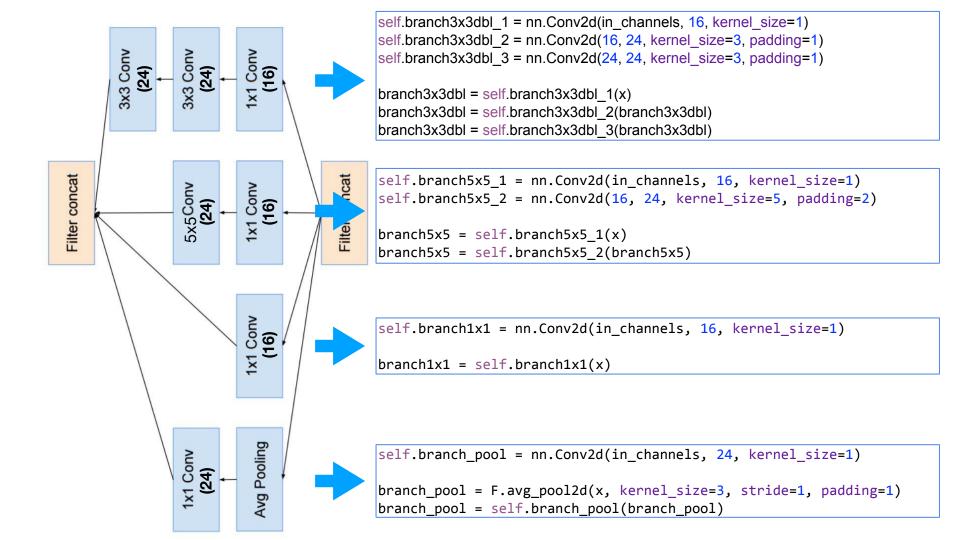


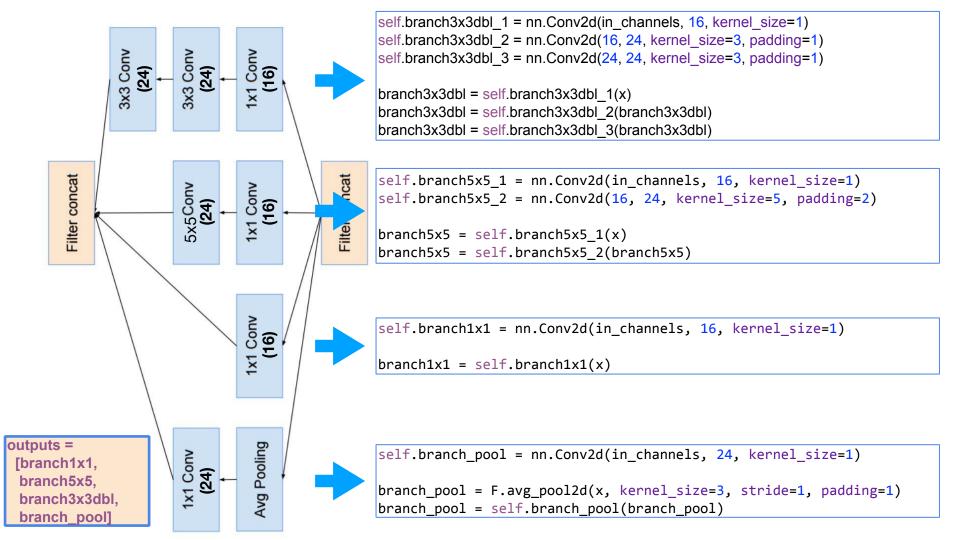












Inception Module

```
| 1x1 Conv (24) | 1x1 Conv (24) | 1x1 Conv (16) | 1x1 Conv (16
```

```
class InceptionA(nn.Module):
   def init (self, in channels):
       super(InceptionA, self). init ()
       self.branch1x1 = nn.Conv2d(in channels, 16, kernel size=1)
       self.branch5x5 1 = nn.Conv2d(in channels, 16, kernel size=1)
       self.branch5x5 2 = nn.Conv2d(16, 24, kernel size=5, padding=2)
       self.branch3x3dbl 1 = nn.Conv2d(in channels, 16, kernel size=1)
       self.branch3x3dbl 2 = nn.Conv2d(16, 24, kernel size=3, padding=1)
       self.branch3x3dbl 3 = nn.Conv2d(24, 24, kernel size=3, padding=1)
       self.branch pool = nn.Conv2d(in channels, 24, kernel size=1)
   def forward(self, x):
       branch1x1 = self.branch1x1(x)
       branch5x5 = self.branch5x5 1(x)
       branch5x5 = self.branch5x5 2(branch5x5)
       branch3x3dbl = self.branch3x3dbl 1(x)
       branch3x3dbl = self.branch3x3dbl 2(branch3x3dbl)
       branch3x3dbl = self.branch3x3dbl 3(branch3x3dbl)
       branch pool = F.avg pool2d(x, kernel size=3, stride=1, padding=1)
       branch pool = self.branch pool(branch pool)
       outputs = [branch1x1, branch5x5, branch3x3dbl, branch pool]
       return torch.cat(outputs, 1)
```

```
class Net(nn.Module):
   def init (self):
       super(Net, self). init ()
       self.conv1 = nn.Conv2d(1, 10, kernel size=5)
       self.conv2 = nn.Conv2d(88, 20, kernel size=5)
       self.incept1 = InceptionA(in channels=10)
       self.incept2 = InceptionA(in channels=20)
       self.mp = nn.MaxPool2d(2)
       self.fc = nn.Linear(1408, 10)
   def forward(self, x):
       in size = x.size(0)
       x = F.relu(self.mp(self.conv1(x)))
       x = self.incept1(x)
       x = F.relu(self.mp(self.conv2(x)))
       x = self.incept2(x)
       x = x.view(in size, -1) # flatten the tensor
       x = self.fc(x)
       return F.log softmax(x)
```

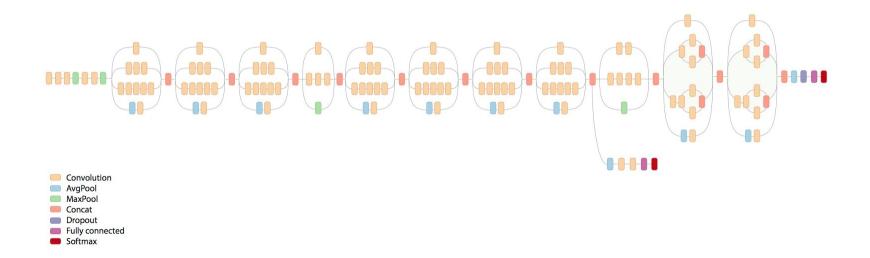
Inception Module

```
Train Epoch: 9 [44800/60000 (75%)]
                                        Loss: 0.064180
Train Epoch: 9 [45440/60000 (76%)]
                                        Loss: 0.020339
Train Epoch: 9 [46080/60000 (77%)]
                                        Loss: 0.061476
Train Epoch: 9 [46720/60000 (78%)]
                                        Loss: 0.039662
Train Epoch: 9 [47360/60000 (79%)]
                                        Loss: 0.026798
Train Epoch: 9 [48000/60000 (80%)]
                                        Loss: 0.071569
Train Epoch: 9 [48640/60000 (81%)]
                                        Loss: 0.003835
Train Epoch: 9 [49280/60000 (82%)]
                                        Loss: 0.005564
Train Epoch: 9 [49920/60000 (83%)]
                                        Loss: 0.020116
Train Epoch: 9 [50560/60000 (84%)]
                                        Loss: 0.128114
Train Epoch: 9 [51200/60000 (85%)]
                                        Loss: 0.016599
                                        Loss: 0.006995
Train Epoch: 9 [51840/60000 (86%)]
Train Epoch: 9 [52480/60000 (87%)]
                                        Loss: 0.111267
Train Epoch: 9 [53120/60000 (88%)]
                                        Loss: 0.052126
Train Epoch: 9 [53760/60000 (90%)]
                                        Loss: 0.034962
Train Epoch: 9 [54400/60000 (91%)]
                                        Loss: 0.029465
Train Epoch: 9 [55040/60000 (92%)]
                                        Loss: 0.031482
Train Epoch: 9 [55680/60000 (93%)]
                                        Loss: 0.015132
Train Epoch: 9 [56320/60000 (94%)]
                                        Loss: 0.010435
Train Epoch: 9 [56960/60000 (95%)]
                                        Loss: 0.014344
Train Epoch: 9 [57600/60000 (96%)]
                                        Loss: 0.014952
Train Epoch: 9 [58240/60000 (97%)]
                                        Loss: 0.153132
Train Epoch: 9 [58880/60000 (98%)]
                                        Loss: 0.112024
Train Epoch: 9 [59520/60000 (99%)]
                                        Loss: 0.009406
```

Test set: Average loss: 0.0470, Accuracy: 9866/10000 (99%)

```
class Net(nn.Module):
   def init (self):
       super(Net, self). init ()
       self.conv1 = nn.Conv2d(1, 10, kernel size=5)
       self.conv2 = nn.Conv2d(88, 20, kernel size=5)
       self.incept1 = InceptionA(in channels=10)
       self.incept2 = InceptionA(in channels=20)
       self.mp = nn.MaxPool2d(2)
       self.fc = nn.Linear(1408, 10)
   def forward(self, x):
       in size = x.size(0)
       x = F.relu(self.mp(self.conv1(x)))
       x = self.incept1(x)
       x = F.relu(self.mp(self.conv2(x)))
       x = self.incept2(x)
       x = x.view(in size, -1) # flatten the tensor
       x = self.fc(x)
       return F.log softmax(x)
```

Exercise 10-2: Implement full inception v3/v4



Exercise 10-3: Implement DenseNet

