



[Return to "Computer Vision Nanodegree" in the classroom](#)

[DISCUSS ON STUDENT HUB](#)

Facial Keypoint Detection

REVIEW

CODE REVIEW 1

HISTORY

▼ models.py 1

```
1 ## TODO: define the convolutional neural network architecture
2
3 import torch
4 import torch.nn as nn
5 import torch.nn.functional as F
6 # can use the below import should you choose to initialize the weights of your
7 import torch.nn.init as I
8
9
10 class Net(nn.Module):
11
12     def __init__(self):
13         super(Net, self).__init__()
14
15         ## TODO: Define all the layers of this CNN, the only requirements are:
16         ## 1. This network takes in a square (same width and height), grayscale
17         ## 2. It ends with a linear layer that represents the keypoints
18         ## it's suggested that you make this last layer output 136 values, 2 for
19
20         # As an example, you've been given a convolutional layer, which you may
21         # 1 input image channel (grayscale), 32 output channels/feature maps, !
22
23         self.conv1 = nn.Conv2d(1, 32, 5)
24         # output size = (W-F)/S + 1 = (224-5)/1 + 1 = 220
25         self.pool1 = nn.MaxPool2d(2, 2)
26         # 220/2 = 110 the output Tensor for one image, will have the #dimension
```

```

27     self.conv2 = nn.Conv2d(32,64,3)
28     # output size = (W-F)/S +1 = (110-3)/1 + 1 = 108
29     self.pool2 = nn.MaxPool2d(2, 2)

```

SUGGESTION

Max pooling is a simple down-sampling operation and I am sure you know what it does. And `nn.MaxPool2d` do not have to declare multiple instances of it, you can just reuse the same object multiple times. 😊

Same goes for the dropout function (for a given dropping probability, of course).

```

30     #108/2=54    the output Tensor for one image, will have the #dimensions
31     self.conv3 = nn.Conv2d(64,128,3)
32     # output size = (W-F)/S +1 = (54-3)/1 + 1 = 52
33     self.pool3 = nn.MaxPool2d(2, 2)
34     #52/2=26    the output Tensor for one image, will have the #dimensions
35     self.conv4 = nn.Conv2d(128,256,3)
36     # output size = (W-F)/S +1 = (26-3)/1 + 1 = 24
37     self.pool4 = nn.MaxPool2d(2, 2)
38     #24/2=12    the output Tensor for one image, will have the #dimensions:
39     self.conv5 = nn.Conv2d(256,512,1)
40     # output size = (W-F)/S +1 = (12-1)/1 + 1 = 12
41     self.pool5 = nn.MaxPool2d(2, 2)
42     #12/2=6     the output Tensor for one image, will have the #dimensions:
43     #Linear Layer
44     self.fc1 = nn.Linear(512*6*6, 1024)
45     self.fc2 = nn.Linear(1024, 136)
46     #self.fc3 = nn.Linear(136, 136)
47     # dropouts
48     self.drop1 = nn.Dropout(p = 0.3)
49     self.drop2 = nn.Dropout(p = 0.3)
50     self.drop3 = nn.Dropout(p = 0.3)
51     self.drop4 = nn.Dropout(p = 0.4)
52     self.drop5 = nn.Dropout(p = 0.3)
53     self.drop6 = nn.Dropout(p = 0.4)
54     #self.drop7 = nn.Dropout(p = 0.5)
55
56
57     ## Note that among the layers to add, consider including:
58     # maxpooling layers, multiple conv layers, fully-connected layers, and
59     # Check out list of layers in pyTorch: batch norm
60
61
62
63     def forward(self, x):
64         ## TODO: Define the feedforward behavior of this model
65         ## x is the input image and, as an example, here you may choose to inc
66         x = self.pool1(F.relu(self.conv1(x)))
67         x = self.drop1(x)
68         x = self.pool2(F.relu(self.conv2(x)))
69         x = self.drop2(x)
70         x = self.pool3(F.relu(self.conv3(x)))
71         x = self.drop3(x)
72         x = self.pool4(F.relu(self.conv4(x)))
73         x = self.drop4(x)
74         x = self.pool5(F.relu(self.conv5(x)))
75         x = self.drop5(x)
76         x = x.view(x.size(0), -1)

```

```
77         x = F.relu(self.fc1(x))
78         x = self.drop6(x)
79         x = self.fc2(x)
80         #x = self.drop7(x)
81         #x = self.fc3(x)
82         # a modified x, having gone through all the layers of your model, should
83         return x
84
85         # a modified x, having gone through all the layers of your model, should
86         #return x
87
```

► detector_architectures/haarcascade_smile.xml

► workspace_utils.py

► filelist.txt

► data_load.py

RETURN TO PATH
