`Super Pixel Pooling` in pytorch

`Super Pixel Pooling` in pytorch - PyTorch Forums

zeakey KAI ZHAO Nov '17

I'm now implementing a pooling layer similar to 'super-pixel Pooling' which has pre-computed superpixel masks to guide the pooling.

Firstly I read the document about extending pytorch which says

You can extend it in both ways, but we recommend using modules for all kinds of layers, that hold any parameters or buffers, and recommend using a functional form parameter-less operations like activation functions, pooling, etc.

So I got confused here, because the operator I want to implement has no *parameter*, but do have a buffer (the superpixel masks to guide the pooling).

Now I just finished my implementation based on 'Module', by inheriting the nn.Module and define the forward():

```
class SuperPixelPooling(nn.Module):
    def __init__(self. input, mask, output):
        super(SuperPixelPooling, self).__init__()
    def forward(self, input, mask)
        # perform superpixel pooling on input according to mask
        output = do_superpixel_pooling(input, mask)
    return output
```

All the operations in forward() are based on Variables.

Another problem occurs: when I'm using the new 'SuperpixelPooling' module in my model, it seems impossible to put such module with multiple inputs into my model.

```
class LeNet(nn.Module):
 def __init__(self):
   super(LeNet, self).__init__()
    self.conv1 = nn.Conv2d(1, 10, kernel size=5)
   self.conv2 = nn.Conv2d(10, 20, kernel_size=5)
   # seems not right?###########
   self.sp_pool2 = SuperPixelPooling()
   # ####################################
   self.conv2_drop = nn.Dropout2d()
    self.fc1 = nn.Linear(320, 50)
   self.fc2 = nn.Linear(50, 10)
 def forward(self, x):
   x = F.relu(F.max_pool2d(self.conv1(x), 2))
   x = F.relu(F.max_pool2d(self.conv2_drop(self.conv2(x)), 2))
   # or like this ? #############
   x = SuperPixelPooling(x, mask)
   #####################################
   x = x.view(-1, 320)
   x = F.relu(self.fc1(x))
   x = F.dropout(x, training=self.training)
   x = self.fc2(x)
   return F.log_softmax(x)
```

I have to use this module with multiple inputs as a Function in my model's forward(), right?

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chenyuntc Yun Chen Nov '17

where is your forward? Do you want something like this:

```
class SuperpixelPooling(nn.Module):
    def __init__(self. input, mask, output):
        super(SuperpixelPooling, self).__init__()
        self.register_buffer('mask', torch.zeros([2222]))
        # do nothing
        # return nothing

    def forward(self, input):
        mask = self.mask
        output,new_mask = do_superpixel_pooling(input,mask)
        self.mask = new_mask
        return output
```

zeakey KAI ZHAO Nov '17

Hi Yun,

Yeah you just got what I want to do. I have updated the code in the question.

It seems impossible to put layers with multiple inputs, just like my SuperPixelPooling into my model with nn.Module, right?

I have to implement the Function version and plug it into my model with super_pixel_pooling = SuperPixelPoolingFunction(input, mask), right?

chenyuntc Yun Chen Nov '17

Of course module could have multi inputs.

```
zeakey:
x = SuperPixelPooling(x, mask)
```

it should be

```
x = self.sp_pool2(x,mask)
```

zeakey KAI ZHAO Nov '17

OK, I will try again.

Thank you!

zeakey KAI ZHAO Nov '17

Hi Chen Yun:

Now I have rewritten SuperPixelPooling operator and it seems work (I got reasonable training loss and testing accuracy),

but I'm not sure whether it is really perfect.

I'm a caffe user before, where I have to write both the forward() and backward() for a new layer.

Here I only implement the forward() for the SuperPixelPooling operator, all the operations in forward() are based on torch.autograd.Variable.

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So do I need to write backward() for the new operator?

chenyuntc Yun Chen

Nov '17

No you don't need it

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