```
1 import torch
2 import torch.nn as nn
3 import torch.nn.functional as F
4 import torch.utils.model_zoo as model_zoo
5 from torch.autograd.variable import Variable
6 from collections import OrderedDict
7 import re
8 import pdb
  __all__ = ['DenseNet', 'densenet121', 'densenet169', 'densenet169_par',
  'densenet201', 'densenet161']
11
12
13 model_urls = {
       'densenet121':
   'https://download.pytorch.org/models/densenet121-a639ec97.pth',
       'densenet169':
   'https://download.pytorch.org/models/densenet169-b2777c0a.pth',
       'densenet201':
16
   'https://download.pytorch.org/models/densenet201-c1103571.pth',
17
       'densenet161':
  'https://download.pytorch.org/models/densenet161-8d451a50.pth',
18 }
19
20
21 def densenet121(pretrained=False, **kwargs):
      r"""Densenet-121 model from
22
       "Densely Connected Convolutional Networks"
23
  <https://arxiv.org/pdf/1608.06993.pdf>`_
24
25
      Aras:
          pretrained (bool): If True, returns a model pre-trained on ImageNet
26
27
28
      model = DenseNet(num_init_features=64, growth_rate=32,
  block_config=(6, 12, 24, 16),
29
                        **kwargs)
30
      if pretrained:
           # '.'s are no longer allowed in module names, but pervious
31
   _DenseLayer
           # has keys 'norm.1', 'relu.1', 'conv.1', 'norm.2', 'relu.2',
   'conv.2'.
           # They are also in the checkpoints in model_urls. This pattern is
33
  used
           # to find such keys.
34
           pattern = re.compile(
35
36
  r'^(.*denselayer\d+\.(?:norm|relu|conv))\.((?:[12])\.(?:weight|bias|running
  _mean|running_var))$')
           state_dict = model_zoo.load_url(model_urls['densenet121'])
37
           for key in list(state_dict.keys()):
38
39
               res = pattern.match(key)
               if res:
40
                   new_key = res.group(1) + res.group(2)
41
42
                   state_dict[new_key] = state_dict[key]
                   del state_dict[key]
43
44
           model.load state dict(state dict)
      model.classifier = None
45
      features = model.features
46
      features.block0 = nn.Sequential(features.conv0, features.norm0,
47
  features.relu0)
```

```
48
49
      features.denseblock1 = nn.Sequential(*list(features.denseblock1))
      features.transition1 = nn.Sequential(*list(features.transition1)[:-1])
50
51
      features.denseblock2 = nn.Sequential(*list(features.denseblock2))
52
53
      features.transition2 = nn.Sequential(*list(features.transition2)[:-1])
54
55
      features.denseblock3 = nn.Sequential(*list(features.denseblock3))
      features.transition3 = nn.Sequential(*list(features.transition3)[:-1])
56
57
      features.denseblock4 =
58
  nn.Sequential(*(list(features.denseblock4)+[features.norm5]))
      model.features = features
59
60
      return model
61
62
  def densenet169_par(pretrained=False, **kwargs):
63
      r"""Densenet-169 model from
64
       "Densely Connected Convolutional Networks"
65
  <https://arxiv.org/pdf/1608.06993.pdf>`_
66
67
      Args:
           pretrained (bool): If True, returns a model pre-trained on ImageNet
68
69
70
      model = DenseNetPar(num_init_features=64, growth_rate=32,
  block_config=(6, 12, 32, 32),
                        **kwarqs)
71
72
      if pretrained:
           # '.'s are no longer allowed in module names, but pervious
  _DenseLayer
           # has keys 'norm.1', 'relu.1', 'conv.1', 'norm.2', 'relu.2',
74
   'conv.2'.
75
           # They are also in the checkpoints in model_urls. This pattern is
  used
76
           # to find such keys.
77
           pattern = re.compile(
78
  r'^(.*denselayer\d+\.(?:norm|relu|conv))\.((?:[12])\.(?:weight|bias|running)
  _mean|running_var))$')
           state dict = model zoo.load url(model urls['densenet169'])
79
           for key in list(state_dict.keys()):
80
               res = pattern.match(key)
81
               if res:
82
83
                   new_key = res.group(1) + res.group(2)
                   state_dict[new_key] = state_dict[key]
84
                   del state_dict[key]
85
86
           model.load_state_dict(state_dict)
      model.classifier = None
87
      features = model.features
88
      features.block0 = nn.Sequential(features.conv0, features.norm0,
  features.relu0)
90
      features.denseblock1 = nn.Sequential(*list(features.denseblock1))
91
92
      features.transition1 = nn.Sequential(*list(features.transition1)[:-1])
93
      features.denseblock2 = nn.Sequential(*list(features.denseblock2))
94
95
      features.transition2 = nn.Sequential(*list(features.transition2)[:-1])
96
      features.denseblock3 = nn.Sequential(*list(features.denseblock3))
97
98
      features.transition3 = nn.Sequential(*list(features.transition3)[:-1])
```

```
99
100
       features.denseblock4 = nn.Sequential(*(list(features.denseblock4)))
       model.features = features
101
102
       return model
103
104
105
   def densenet169(pretrained=False, **kwargs):
106
       r"""Densenet-169 model from
107
       "Densely Connected Convolutional Networks"
   <https://arxiv.org/pdf/1608.06993.pdf>`_
108
109
       Args:
           pretrained (bool): If True, returns a model pre-trained on ImageNet
110
111
112
       model = DenseNet(num_init_features=64, growth_rate=32,
   block_config=(6, 12, 32, 32),
                         **kwargs)
113
       if pretrained:
114
           # '.'s are no longer allowed in module names, but pervious
115
   _DenseLayer
           # has keys 'norm.1', 'relu.1', 'conv.1', 'norm.2', 'relu.2',
116
   'conv.2'.
           # They are also in the checkpoints in model_urls. This pattern is
117
   used
118
           # to find such keys.
           pattern = re.compile(
119
120
   r'^{(.*denselayer\d+\.(?:norm|relu|conv))}.((?:[12])\.(?:weight|bias|running)
   _mean|running_var))$')
           state_dict = model_zoo.load_url(model_urls['densenet169'])
121
           for key in list(state_dict.keys()):
122
                res = pattern.match(key)
123
124
                if res:
125
                    new_key = res.group(1) + res.group(2)
126
                    state_dict[new_key] = state_dict[key]
                    del state_dict[key]
127
           model.load_state_dict(state_dict)
128
       model.classifier = None
129
       features = model.features
130
       features.block0 = nn.Sequential(features.conv0, features.norm0,
   features.relu0, features.pool0)
132
       features.denseblock1 = nn.Sequential(*list(features.denseblock1))
133
134
       features.transition1 = nn.Sequential(*list(features.transition1))
135
       features.denseblock2 = nn.Sequential(*list(features.denseblock2))
136
       features.transition2 = nn.Sequential(*list(features.transition2))
137
138
139
       features.denseblock3 = nn.Sequential(*list(features.denseblock3))
       features.transition3 = nn.Sequential(*list(features.transition3))
140
141
       features.denseblock4 = nn.Sequential(*(list(features.denseblock4) +
142
   [features.norm5]))
143
       model.features = features
       return model
144
145
146
   def densenet201(pretrained=False, **kwargs):
147
       r"""Densenet-201 model from
148
       `"Densely Connected Convolutional Networks"
149
```

```
<https://arxiv.org/pdf/1608.06993.pdf>`_
150
151
       Args:
           pretrained (bool): If True, returns a model pre-trained on ImageNet
152
153
154
       model = DenseNet(num_init_features=64, growth_rate=32,
   block\_config=(6, 12, 48, 32),
155
                         **kwarqs)
       if pretrained:
156
            # '.'s are no longer allowed in module names, but pervious
157
   _DenseLayer
            # has keys 'norm.1', 'relu.1', 'conv.1', 'norm.2', 'relu.2',
158
   'conv.2'.
159
            # They are also in the checkpoints in model_urls. This pattern is
   used
160
            # to find such keys.
161
            pattern = re.compile(
   r'^(.*denselayer\d+\.(?:norm|relu|conv))\.((?:[12])\.(?:weight|bias|running)
   _mean|running_var))$')
163
            state_dict = model_zoo.load_url(model_urls['densenet201'])
164
            for key in list(state_dict.keys()):
165
                res = pattern.match(key)
                if res:
166
167
                    new_key = res.group(1) + res.group(2)
                    state_dict[new_key] = state_dict[key]
168
                    del state_dict[key]
169
170
            model.load state dict(state dict)
171
       model.classifier = None
       features = model.features
172
       features.block0 = nn.Sequential(features.conv0, features.norm0,
173
   features.relu0, features.pool0)
174
175
       features.denseblock1 = nn.Sequential(*list(features.denseblock1))
176
       features.transition1 = nn.Sequential(*list(features.transition1))
177
178
       features.denseblock2 = nn.Sequential(*list(features.denseblock2))
       features.transition2 = nn.Sequential(*list(features.transition2))
179
180
       features.denseblock3 = nn.Sequential(*list(features.denseblock3))
181
       features.transition3 = nn.Sequential(*list(features.transition3))
182
183
       features.denseblock4 = nn.Sequential(*(list(features.denseblock4) +
184
   [features.norm5]))
       model.features = features
185
       return model
186
187
188
   def densenet161(pretrained=False, **kwargs):
189
       r"""Densenet-161 model from
190
        "Densely Connected Convolutional Networks"
191
   <https://arxiv.org/pdf/1608.06993.pdf>`_
192
193
       Args:
           pretrained (bool): If True, returns a model pre-trained on ImageNet
194
195
196
       model = DenseNet(num_init_features=96, growth_rate=48,
   block_config=(6, 12, 36, 24),
                         **kwargs)
197
198
       if pretrained:
```

```
# '.'s are no longer allowed in module names, but pervious
199
   DenseLaver
           # has keys 'norm.1', 'relu.1', 'conv.1', 'norm.2', 'relu.2',
   'conv.2'.
201
           # They are also in the checkpoints in model_urls. This pattern is
   used
202
           # to find such keys.
203
           pattern = re.compile(
204
   r'^(.*denselayer\d+\.(?:norm|relu|conv))\.((?:[12])\.(?:weight|bias|running))
   _mean|running_var))$')
           state dict = model zoo.load url(model urls['densenet161'])
205
            for key in list(state_dict.keys()):
206
207
                res = pattern.match(key)
208
                if res:
209
                    new_key = res.group(1) + res.group(2)
210
                    state_dict[new_key] = state_dict[key]
211
                    del state_dict[key]
212
           model.load state dict(state dict)
       model.classifier = None
213
214
       features = model.features
215
       features.block0 = nn.Sequential(features.conv0, features.norm0,
   features.relu0)
216
217
       features.denseblock1 = nn.Sequential(*list(features.denseblock1))
218
       features.transition1 = nn.Sequential(*list(features.transition1)[:-1])
219
220
       features.denseblock2 = nn.Sequential(*list(features.denseblock2))
221
       features.transition2 = nn.Sequential(*list(features.transition2)[:-1])
222
       features.denseblock3 = nn.Sequential(*list(features.denseblock3))
223
       features.transition3 = nn.Sequential(*list(features.transition3)[:-1])
224
225
226
       features.denseblock4 = nn.Sequential(*(list(features.denseblock4) +
   [features.norm5]))
       model.features = features
227
       return model
228
229
230
   class DenseLayer(nn.Sequential):
231
       def __init__(self, num_input_features, growth_rate, bn_size,
232
   drop_rate, dilation):
233
           super(_DenseLayer, self).__init__()
234
           self.add_module('norm1', nn.BatchNorm2d(num_input_features)),
           self.add_module('relu1', nn.ReLU(inplace=True)),
235
           self.add_module('conv1', nn.Conv2d(num_input_features, bn_size *
236
237
                            growth_rate, kernel_size=1, stride=1, bias=False)),
           self.add_module('norm2', nn.BatchNorm2d(bn_size * growth_rate)),
238
           self.add_module('relu2', nn.ReLU(inplace=True)),
239
           self.add_module('conv2', nn.Conv2d(bn_size * growth_rate,
   growth_rate,
241
                            kernel_size=3, stride=1, padding=dilation,
   bias=False, dilation=dilation)),
242
           self.drop_rate = drop_rate
243
244
       def forward(self, x):
           new_features = super(_DenseLayer, self).forward(x)
245
246
           if self.drop_rate > 0:
                new_features = F.dropout(new_features, p=self.drop_rate,
   training=self.training)
```

```
return torch.cat([x, new_features], 1)
248
249
250
251
   class _DenseBlock(nn.Sequential):
       def __init__(self, num_layers, num_input_features, bn_size,
252
   growth_rate, drop_rate, dilation=1):
253
           super(_DenseBlock, self).__init__()
254
           for i in range(num_layers):
255
                layer = _DenseLayer(num_input_features + i * growth_rate,
   growth_rate, bn_size, drop_rate, dilation)
                self.add_module('denselayer%d' % (i + 1), layer)
256
257
258
259
   class Transition(nn.Sequential):
       def __init__(self, num_input_features, num_output_features):
260
           super(_Transition, self).__init__()
261
           self.add_module('norm', nn.BatchNorm2d(num_input_features))
262
263
           self.add_module('relu', nn.ReLU(inplace=True))
           self.add_module('conv', nn.Conv2d(num_input_features,
264
   num_output_features,
265
                                               kernel_size=1, stride=1,
   bias=False))
266
           self.add_module('pool', nn.AvgPool2d(kernel_size=2, stride=2))
267
268
269 class DenseNet(nn.Module):
       r"""Densenet-BC model class, based on
270
        "Densely Connected Convolutional Networks"
271
   <https://arxiv.org/pdf/1608.06993.pdf>`_
272
273
       Aras:
           growth_rate (int) - how many filters to add each layer (`k` in
274
275
           block_config (list of 4 ints) - how many layers in each pooling
   block
           num_init_features (int) - the number of filters to learn in the
276
   first convolution layer
277
           bn_size (int) - multiplicative factor for number of bottle neck
   layers
              (i.e. bn size * k features in the bottleneck layer)
278
279
           drop_rate (float) - dropout rate after each dense layer
           num_classes (int) - number of classification classes
280
281
282
       def __init__(self, growth_rate=32, block_config=(6, 12, 24, 16),
                     num_init_features=64, bn_size=4, drop_rate=0,
283
   num_classes=1000,
                     is_downsamples=[True, True, True, True], dilations=[1,
284
   1, 1, 1, 1]):
285
           super(DenseNet, self).__init__()
286
287
           self.is_downsamples = is_downsamples
           self.dilations = dilations
288
           # First convolution
289
290
           self.features = nn.Sequential(OrderedDict([
                ('conv0', nn.Conv2d(3, num_init_features, kernel_size=7,
291
   stride=2, padding=3, bias=False)),
292
                ('norm0', nn.BatchNorm2d(num_init_features)),
293
                ('relu0', nn.ReLU(inplace=True)),
                ('pool0', nn.MaxPool2d(kernel_size=3, stride=2, padding=1)),
294
295
            ]))
```

```
296
297
            # Each denseblock
298
            num features = num init features
299
            for i, num_layers in enumerate(block_config):
300
                block = _DenseBlock(num_layers=num_layers,
   num_input_features=num_features,
301
                                     bn_size=bn_size, growth_rate=growth_rate,
   drop_rate=drop_rate, dilation=dilations[i])
302
                self.features.add_module('denseblock%d' % (i + 1), block)
                num_features = num_features + num_layers * growth_rate
303
                if i != len(block_config) - 1:
304
305
                    trans = _Transition(num_input_features=num_features,
   num_output_features=num_features // 2)
                    self.features.add_module('transition%d' % (i + 1), trans)
306
307
                    num_features = num_features // 2
308
309
            # Final batch norm
310
            self.features.add_module('norm5', nn.BatchNorm2d(num_features))
311
312
            # Linear layer
313
            self.classifier = nn.Linear(num_features, num_classes)
314
315
            # Official init from torch repo.
            for m in self.modules():
316
317
                if isinstance(m, nn.Conv2d):
318
                    nn.init.kaiming_normal(m.weight.data)
319
                elif isinstance(m, nn.BatchNorm2d):
320
                    m.weight.data.fill_(1)
321
                    m.bias.data.zero_()
322
                elif isinstance(m, nn.Linear):
323
                    m.bias.data.zero_()
324
325
       def forward(self, x):
326
            x = self.features.block0(x)
327
            x = self.features.denseblock1(x)
328
           x = self.features.transition1(x)
329
330
331
            x = self.features.denseblock2(x)
            x = self.features.transition2(x)
332
333
            x = self.features.denseblock3(x)
334
            x = self.features.transition3(x)
335
336
            x = self.features.denseblock4(x)
337
338
            return x
339
340
341 if __name__ == "__main__":
       net = densenet169(pretrained=True).cuda()
342
343
       pdb.set_trace()
       x = torch.Tensor(2, 3, 256, 256).cuda()
344
345
       sb = net(Variable(x))
346
       pdb.set_trace()
```