

[< Back to AI Programming with Python Nanodegree](#)

# Image Classifier Application

REVIEW

CODE REVIEW 8

HISTORY

▶ train.py 4

▼ predict.py 3

```

1 #!/usr/bin/env python3
2 # -*- coding: utf-8 -*-
3 # *AIPND/ImageClassifierApp/predict.py
4
5 #
6 # reads in an image and a checkpoint then prints the most likely image class
7 # and it's associated probability.
8 # specs :
9 # The script reads in an image and a checkpoint to print the most likely image class
10 # Allow user to load a JSON file that maps the class values to other category names
11 # Allow user to request a print out of top k classes and probabilities
12 # Allow user to request the predictions be done in GPU model
13 #
14 # About command line arguments of predict.py :
15 #   checkpoint : specify a saved checkpoint in dir 'chksav'. If not supplied, use 'chksav/chkpt.pth'
16 #   --img_pth : specify an image in dir 'flowers'. If not supplied, use 'flowers/test/91/image_08061.jpg'
17 #   --category_names : specify a category name JSON mapper file. If not supplied, use 'cat_name_mapper.json'
18 #   --top_k : specify no. of top k classes to print. Default is 1
19 #   --gpu : run predict in GPU mode (subject to device capability), default is 0
20 # Example Calls
21 # Ex 1, use checkpoint 'chkpt.pth' in 'chksav': python predict.py chksav/chkpt.pth
22 # Ex 2, use top_k 4 and GPU : python predict.py --top_k 4 --gpu
23 # Ex 3, use img_pth 'flowers/test/91/image_08061.jpg' and cat name mapper 'cat_name_mapper.json'
24 #   python predict.py --img_pth flowers/test/91/image_08061.jpg --category_names:

```

```

25 #
26 import argparse
27 import torch
28 from torchvision import models
29 import numpy as np
30 from PIL import Image
31
32 import json
33 from datetime import datetime
34 import os
35 import glob
36 import sys
37
38 from workspace_utils import active_session
39
40 device = torch.device('cuda:0' if torch.cuda.is_available() else 'cpu')
41 data_dir = 'flowers'
42 test_dir = data_dir + '/test'
43 chkptdir = 'chksav'
44 # use the last checkpoint saved as default
45 if len(glob.glob(chkptdir+'/*.pth')) > 0 :
46     chkpt = max(glob.glob(chkptdir+'/*.pth'), key=os.path.getctime)
47 else:
48     chkpt = None
49     print('\n*** no saved checkpoint to load ... exiting\n')
50     sys.exit(1)
51
52 def main():
53     # collecting start time
54     start_time = datetime.now()
55
56     # get input arguments
57     args = get_input_args()
58     print('\n*** command line arguments ***')
59     print('checkpoint:', args.checkpoint, '\nimage path:', args.img_pth,
60           '\ncategory names mapper file:', args.category_names, '\nno. of top k
61           '\nGPU mode:', args.gpu, '\n')
62
63     if len(glob.glob(args.checkpoint)) == 0:
64         print('*** checkpoint: ', args.checkpoint, ', not found ... exiting\n')
65         sys.exit(1)
66
67     if len(glob.glob(args.img_pth)) == 0:
68         print('*** img_pth: ', args.img_pth, ', not found ... exiting\n')
69         sys.exit(1)
70
71     if len(glob.glob(args.category_names)) == 0 :
72         print('*** category names mapper file: ', args.category_names, ', not fo
73         sys.exit(1)
74
75     if args.top_k < 1:
76         print('*** no. of top k classes to print must >= 1 ... exiting\n')
77         sys.exit(1)
78
79     if device.type == 'cuda':
80         if args.gpu:
81             print('*** GPU is available, using GPU ...\n')
82         else:
83             print('*** using GPU ...\n')
84     else:

```

```

85         if args.gpu:
86             print('*** GPU is unavailable, using CPU ...\n')
87         else:
88             print('*** using CPU ...\n')
89
90     # retrieve model from checkpoint saved
91     model = load_checkpoint(args)
92
93     # call predic function with required and optional input parameters
94     with active_session():
95         predict(model, args)
96
97     elapsed = datetime.now() - start_time
98     print('\n*** prediction done ! \nElapsed time[hh:mm:ss.ms]: {}'.format(elapsed))
99
100 def get_input_args():
101     # create parser
102     parser = argparse.ArgumentParser()
103
104     parser.add_argument('checkpoint', type=str, nargs='?', default=checkpoint,
105                         help='path to saved checkpoint')
106
107     parser.add_argument('-img', '--img_pth', type=str, default=test_dir + '/91/imgs',
108                         help='path to an image file')
109
110     parser.add_argument('-cat', '--category_names', dest='category_names', default=category_names,
111                         type=str, help='path to JSON file for mapping class values')
112
113     parser.add_argument('-k', '--top_k', dest='top_k', default=1, type=int,
114                         help='no. of top k classes to print')
115
116     parser.add_argument('--gpu', dest='gpu', default=False, action='store_true',
117                         help='predict in gpu mode')
118
119     return parser.parse_args()
120
121 def load_checkpoint(args):
122     if device.type == 'cuda':
123         print('*** loading chkpt', args.checkpoint, 'in cuda ...\n')
124         checkpoint = torch.load(args.checkpoint)
125     else:
126         print('*** loading chkpt', args.checkpoint, 'in cpu ...\n')
127         checkpoint = torch.load(args.checkpoint, map_location=lambda storage, loc: storage)
128
129     model = models.__dict__[checkpoint['arch']](pretrained=True)
130     model.classifier = checkpoint['classifier']
131     model.class_to_idx = checkpoint['class_to_idx']
132     model.load_state_dict(checkpoint['state_dict'])
133
134     return model
135
136 def process_image(image):
137     ''' Scales, crops, and normalizes a PIL image for a PyTorch model,
138         returns an Numpy array
139     '''

```



AWESOME

## Great job using docstrings to document major classes and methods!

To learn more about the benefits of documentation, check out [this article](#).

```
140
141     pil_img=image
142
143     sz = image.size
144     h = min(image.size)
145     w = max(image.size)
146     #print('size:',sz, ', h:',h, ', w:',w)
147
148     # calculate ratio_aspect using original height & width
149     # chosen h is 256, ratio aspect for adjusted w is original w/h
150     ratio_aspect = w/h
151
152     # get indices of short and long sides
153     x = image.size.index(min(image.size))
154     y = image.size.index(max(image.size))
155
156     # calc new size with short side 256 pixels keeping ratio aspect
157     new_sz = [0, 0]
158     new_sz[x] = 256
159     new_sz[y] = int(new_sz[x] * ratio_aspect)
160
161     #print('new_sz:',new_sz, '\npre resized img:', pil_img)
162
163     # resize base on short side of 256 pixels
164     pil_img=image.resize(new_sz)
165     #print('post resized image:', pil_img)
166
167     # crop out the center 224x224 portion
168     wid, hgt = new_sz
169     #print('wid:', wid, ', hgt:', hgt)
170
171     # calc left, top, right, bottom margin pos
172     l_margin = (wid - 224)/2
173     t_margin = (hgt - 224)/2
174     r_margin = (wid + 224 )/2
175     b_margin = (hgt + 224)/2
176
177     #print('left:',l_margin, ', top:',t_margin, ', right:',r_margin, ', bottom:
178
179     # crop the image
180     pil_img=pil_img.crop((l_margin, t_margin, r_margin, b_margin))
181     #print('cropped img:', pil_img)
182
183     # convert to np array for normalization purpose
184     np_img = np.array(pil_img)
185
186     print('np_img.shape',np_img.shape)
187
188     np_img = np_img/255
189     mean = np.array([0.485, 0.456, 0.406])
190     std = np.array([0.229, 0.224, 0.225])
191     np_img = (np_img - mean)/std
192
193     # transpose to get color channel to 1st pos
194     np_img = np_img.transpose((2, 0, 1))
```

```

195
196     return np_img
197
198
199 def predict(model, args):
200     ''' Predict the class (or classes) of an image using a trained deep learning
201         '''
202
203     model.cpu()
204     model.eval()
205
206     pil_img = Image.open(args.img_pth)
207     image = process_image(pil_img)
208     image = torch.FloatTensor(image)
209
210     model, image = model.to(device), image.to(device)

```



AWESOME

Nicely done configuring the application to run on GPU!

```

211
212     print('\nori image.shape:', image.shape)
213     image.unsqueeze_(0) # add a new dimension in pos 0
214     print('new image.shape:', image.shape, '\n')
215
216
217     output = model.forward(image)
218
219     # get the top k classes of prob
220     ps = torch.exp(output).data[0]
221     topk_prob, topk_idx = ps.topk(args.top_k)
222
223     # bring back to cpu and convert to numpy
224
225     topk_probs = topk_prob.cpu().numpy()
226     topk_idx = topk_idx.cpu().numpy()
227
228     # map topk_idx to classes in model.class_to_idx
229     idx_class={i: k for k, i in model.class_to_idx.items()}
230     topk_classes = [idx_class[i] for i in topk_idx]
231
232     print('*** Top ', args.top_k, ' classes ***')
233     # map class to class name
234     if args.category_names:
235         with open(args.category_names, 'r') as f:

```



AWESOME

Nicely done using the user specified JSON file for category mapping!

```

236         cat_to_name = json.load(f)
237
238     topk_names = [cat_to_name[i] for i in topk_classes]
239     print('class names: ', topk_names)

```

```
240
241     print('classes:      ', topk_classes)
242     print('probabilities: ', topk_probs)
243
244
245
246 # Call to main function to run the program
247 if __name__ == "__main__":
248     main()
249
```

► README.md 1

► workspace\_utils.py

RETURN TO PATH

Rate this review

---

[Student FAQ](#)