## **Super resolution**

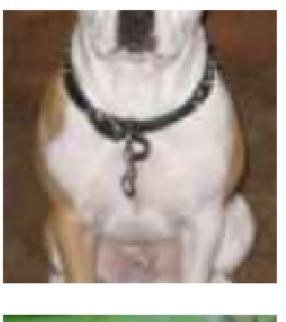
```
In [1]:
             import fastai
             from fastai.vision import *
            from fastai.callbacks import *
            from torchvision.models import vgg16 bn
In [2]:
             import os
            from os import listdir
            from os.path import isfile, join
             import shutil
In [3]:
             # styled files = [styled file.split('style ')[1] for styled file in os.listd
             # unprocessed content = [content img for content img in os.listdir(content i
          2
          3
          4
             # for content img in [content img for content img in unprocessed content if
          5
                   print(content_img)
                   out_img = style_transfer(content_img)
          6
                   out_img.save(output_dir + 'style_' + content_img)
In [4]:
             styled src = Path('../../TampaAI/N A A Style/output/pets monet sunset')
            full size img src dir = '/home/ml1/.fastai/data/oxford-iiit-pet/images/'
          3 img dir = Path('images')
          4 full size dir = img dir/'orig'
          5
            styled_dir = img_dir/'styled'
          6 small 96 dir = img dir/'small-96'
             small 256 dir = img dir/'small-256'
```

```
In [5]:
          1
          2
             styled files = [f for f in listdir(styled src) if isfile(join(styled src, f)
          3 root of styled files = [styled file.split('style ')[1] for styled file in st
             root of styled files
Out[5]: ['great_pyrenees_177.jpg',
          'newfoundland_181.jpg',
          'Bombay 37.jpg',
          'staffordshire bull terrier 137.jpg',
          'japanese_chin_36.jpg',
          'Persian 138.jpg',
          'Ragdoll_90.jpg',
          'scottish_terrier_198.jpg',
          'boxer_197.jpg',
          'Birman 183.jpg',
          'Siamese_30.jpg',
          'keeshond 65.jpg',
          'staffordshire_bull_terrier_73.jpg',
          'Maine_Coon_108.jpg',
          'staffordshire_bull_terrier_107.jpg',
          'english cocker spaniel 117.jpg',
          'pug_170.jpg',
          'Siamese 122.jpg',
          'miniature_pinscher_4.jpg',
             content_img_dir = '/home/ml1/.fastai/data/oxford-iiit-pet/images/'
In [6]:
            # processed content = [content img for content img in os.listdir(content img
             # processed content
In [7]:
             for file name in root of styled files:
          1
          2
                 full file name = os.path.join(full size img src dir, file name)
          3
                 shutil.copy(full file name, full size dir)
In [8]:
             for file_name in root_of_styled_files:
          1
          2
                 full file name = os.path.join(styled src, 'style '+file name)
          3
                 shutil.copy(full_file_name, styled_dir)
                 style_file_name = 'style_' + file_name
          4
          5
                 shutil.move(styled dir/style file name, styled dir/file name)
In [ ]:
In [9]:
             # from pathlib import Path
          1
          2
            # import glob
          3 # # path = untar data(URLs.PETS)
            # styled_src = Path('../..')
          5
             # for file name in glob.glob(styled src):
          6
                   print(file name)
          7
```

```
In [10]:
             # path hr = path/'images'
             # path lr = path/'small-96'
           2
           3 | # path mr = path/'small-256'
             path hr = styled dir
             path lr = small 96 dir
              path mr = small 256 dir
In [11]:
              il = ImageItemList.from_folder(full_size_dir)
              # f list = il.to df().head()
In [12]:
In [13]:
              def resize one(fn,i):
           1
           2
                    dest = path_lr/fn.relative_to(path_hr)
                  dest = path lr/fn.relative_to(full_size_dir)
           3
           4
                  dest.parent.mkdir(parents=True, exist ok=True)
           5
                  img = PIL.Image.open(fn)
                  targ sz = resize to(img, 96, use min=True)
           6
           7
                  img = img.resize(targ sz, resample=PIL.Image.BILINEAR).convert('RGB')
                  img.save(dest, quality=60)
In [14]:
              # to create smaller images, uncomment the next line when you run this the fi
           1
              # parallel(resize one, il.items)
In [15]:
           1
              bs, size=32,128
           2
              arch = models.resnet34
           3
              src = ImageImageList.from folder(path lr).random split by pct(0.1, seed=42)
In [16]:
              src
Out[16]: ItemLists;
         Train: ImageImageList (2753 items)
         [Image (3, 96, 128), Image (3, 96, 128), Image (3, 96, 125), Image (3, 96, 12
         8), Image (3, 96, 128)]...
         Path: images/small-96;
         Valid: ImageImageList (305 items)
         [Image (3, 175, 96), Image (3, 96, 107), Image (3, 96, 125), Image (3, 96, 12
         8), Image (3, 96, 144)]...
         Path: images/small-96;
         Test: None
In [17]:
           1
              def get data(bs,size):
           2
                  data = (src.label from func(lambda x: path hr/x.name)
           3
                          .transform(get_transforms(max_zoom=2.), size=size, tfm_y=True)
           4
                         .databunch(bs=bs).normalize(imagenet stats, do y=True))
           5
           6
                  data.c = 3
           7
                  return data
```

```
In [18]: 1 data = get_data(bs,size)
```

In [19]: 1 data.show\_batch(ds\_type=DatasetType.Valid, rows=2, figsize=(9,9))









## **Feature loss**

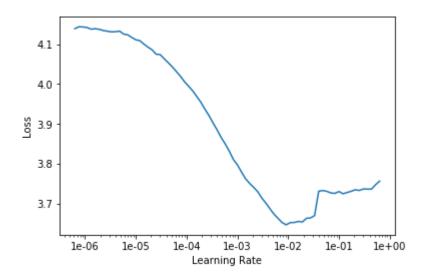
```
In [22]:
              gram matrix(t)
Out[22]: tensor([[[0.1315, 0.1074, 0.0764],
                  [0.1074, 0.0923, 0.0687],
                   [0.0764, 0.0687, 0.0562]],
                  [[0.1315, 0.1074, 0.0764],
                  [0.1074, 0.0923, 0.0687],
                  [0.0764, 0.0687, 0.0562]]])
In [23]:
              base loss = F.ll loss
In [24]:
              vgg m = vgg16 bn(True).features.cuda().eval()
              requires_grad(vgg_m, False)
              blocks = [i-1 for i,o in enumerate(children(vgg m)) if isinstance(o,nn.MaxPo
In [25]:
              blocks, [vgg_m[i] for i in blocks]
Out[25]: ([5, 12, 22, 32, 42],
          [ReLU(inplace), ReLU(inplace), ReLU(inplace), ReLU(inplace)])
In [26]:
           1
              class FeatureLoss(nn.Module):
           2
                  def __init__(self, m_feat, layer_ids, layer_wgts):
                      super().__init__()
           3
           4
                      self.m feat = m feat
           5
                      self.loss features = [self.m feat[i] for i in layer ids]
           6
                      self.hooks = hook_outputs(self.loss_features, detach=False)
           7
                      self.wgts = layer wgts
           8
                      self.metric_names = ['pixel',] + [f'feat_{i}' for i in range(len(lay
           9
                            | + [f'gram {i}' for i in range(len(layer ids))]
          10
          11
                  def make features(self, x, clone=False):
          12
                      self.m feat(x)
          13
                      return [(o.clone() if clone else o) for o in self.hooks.stored]
          14
                  def forward(self, input, target):
          15
                      out feat = self.make features(target, clone=True)
          16
          17
                      in feat = self.make features(input)
                      self.feat_losses = [base_loss(input,target)]
          18
          19
                      self.feat losses += [base loss(f in, f out)*w
                                           for f_in, f_out, w in zip(in_feat, out_feat, se
          20
          21
                      self.feat losses += [base loss(gram matrix(f in), gram matrix(f out)
          22
                                           for f in, f out, w in zip(in feat, out feat, se
          23
                      self.metrics = dict(zip(self.metric names, self.feat losses))
          24
                      return sum(self.feat_losses)
          25
          26
                  def del (self): self.hooks.remove()
In [27]:
              feat_loss = FeatureLoss(vgg_m, blocks[2:5], [5,15,2])
```

## **Train**

Out[28]: 32

```
In [29]: 1 learn.lr_find()
2 learn.recorder.plot()
```

LR Finder is complete, type {learner\_name}.recorder.plot() to see the graph.



In [32]: 1 do\_fit('1a', slice(lr\*10))

Total time: 04:27

epoch	train_loss	valid_loss	pixel	feat_0	feat_1	feat_2	gram_0
1	2.888878	2.802722	0.400150	0.311663	0.346429	0.139423	0.516962
2	2.760338	2.711827	0.380862	0.307217	0.342594	0.137599	0.483044
3	2.710657	2.681858	0.385092	0.307107	0.336578	0.135424	0.471399
4	2.684432	2.710310	0.432943	0.308145	0.332644	0.134307	0.467406
5	2.672602	2.658380	0.393814	0.303303	0.332759	0.132464	0.472350
6	2.645839	2.582036	0.370325	0.300303	0.327458	0.131722	0.448097
7	2.623099	2.554564	0.368303	0.299261	0.327110	0.131516	0.430205
8	2.611051	2.564203	0.379349	0.300568	0.326092	0.130733	0.432082
9	2.586885	2.528744	0.371698	0.299624	0.323526	0.130359	0.421029
10	2.542103	2.467600	0.349594	0.297600	0.321580	0.128988	0.405693

Input / Prediction / Target







In [33]: 1 learn.unfreeze()

In [34]: 1 do\_fit('1b', slice(1e-5,lr))

Total time: 04:38

epoch	train_loss	valid_loss	pixel	feat_0	feat_1	feat_2	gram_0
1	2.508552	2.470243	0.348827	0.298201	0.321827	0.129125	0.405801
2	2.502695	2.466722	0.347024	0.297699	0.321056	0.129063	0.406118
3	2.494808	2.463147	0.347236	0.297211	0.321384	0.129191	0.405461
4	2.498787	2.449929	0.345081	0.297055	0.320709	0.128554	0.400691
5	2.492712	2.469727	0.345163	0.296634	0.320326	0.128987	0.409961
6	2.487676	2.465504	0.346660	0.296866	0.320187	0.129350	0.406975
7	2.476392	2.437048	0.343815	0.295747	0.318131	0.127907	0.397501
8	2.477364	2.429764	0.342185	0.296188	0.318579	0.128037	0.394582
9	2.476491	2.434588	0.340103	0.295186	0.318055	0.127891	0.399896
10	2.466644	2.422903	0.338207	0.295324	0.317424	0.127604	0.394378

Input / Prediction / Target







```
In [35]:
              data = get data(12, size*2)
           1
              data
Out[35]: ImageDataBunch;
         Train: LabelList
         y: ImageItemList (2753 items)
         [Image (3, 256, 342), Image (3, 256, 341), Image (3, 256, 335), Image (3, 256,
         341), Image (3, 256, 341)]...
         Path: images/small-96
         x: ImageImageList (2753 items)
         [Image (3, 96, 128), Image (3, 96, 128), Image (3, 96, 125), Image (3, 96, 12
         8), Image (3, 96, 128)]...
         Path: images/small-96;
         Valid: LabelList
         y: ImageItemList (305 items)
         [Image (3, 468, 256), Image (3, 256, 287), Image (3, 256, 335), Image (3, 256,
         341), Image (3, 256, 384)]...
         Path: images/small-96
         x: ImageImageList (305 items)
         [Image (3, 175, 96), Image (3, 96, 107), Image (3, 96, 125), Image (3, 96, 12
         8), Image (3, 96, 144)]...
         Path: images/small-96;
         Test: None
In [36]:
           1
             learn.data = data
           2 learn.freeze()
              gc.collect()
           3
Out[36]: 19688
In [37]:
              learn
Out[37]: Learner(data=ImageDataBunch;
         Train: LabelList
         y: ImageItemList (2753 items)
         [Image (3, 256, 342), Image (3, 256, 341), Image (3, 256, 335), Image (3, 25
         6, 341), Image (3, 256, 341)]...
         Path: images/small-96
         x: ImageImageList (2753 items)
         [Image (3, 96, 128), Image (3, 96, 128), Image (3, 96, 125), Image (3, 96, 12
         8), Image (3, 96, 128)]...
         Path: images/small-96;
         Valid: LabelList
         y: ImageItemList (305 items)
         [Image (3, 468, 256), Image (3, 256, 287), Image (3, 256, 335), Image (3, 25
         6, 341), Image (3, 256, 384)]...
         Path: images/small-96
         x: ImageImageList (305 items)
         [Image (3, 175, 96), Image (3, 96, 107), Image (3, 96, 125), Image (3, 96, 12
```

```
In [38]:
                learn.data.batch_size
Out[38]: 12
In [39]:
                 learn.data.batch_size = learn.data.batch_size//2
                learn.data.batch_size
Out[39]: 6
In [40]:
                learn.load('1b');
In [41]:
                gc.collect()
Out[41]: 0
                do_fit('2a')
In [42]:
                                                                                                0.258037
                     3
                          1.714106
                                     1.718336
                                                 0.356518
                                                             0.307389
                                                                         0.269749
                                                                                    0.094273
                     4
                          1.688799
                                     1.716100
                                                 0.356088
                                                             0.307136
                                                                         0.269905
                                                                                    0.094825
                                                                                                0.256647
                     5
                          1.684214
                                     1.706422
                                                 0.353547
                                                             0.307031
                                                                         0.268583
                                                                                    0.094448
                                                                                                0.252158
                     6
                                                                                                0.255058
                          1.683811
                                     1.709307
                                                 0.358351
                                                             0.306697
                                                                         0.268672
                                                                                    0.093963
                     7
                          1.670471
                                     1.685609
                                                 0.347387
                                                             0.305498
                                                                         0.267441
                                                                                    0.094377
                                                                                                0.247408
                     8
                                                             0.305030
                                                                                                0.246055
                          1.666559
                                     1.681222
                                                 0.348621
                                                                         0.267201
                                                                                    0.093405
                     9
                          1.668058
                                     1.673358
                                                 0.344755
                                                             0.305114
                                                                         0.268340
                                                                                    0.094490
                                                                                                0.242541
                    10
                          1.641732
                                     1.659688
                                                 0.339444
                                                             0.305622
                                                                                    0.093266
                                                                                                0.239610
                                                                         0.265179
                                              Input / Prediction / Target
                learn.unfreeze()
In [43]:
```

http://ml1:8888/notebooks/scott/TampaAl/repair\_images/superres-pets-monet-sunset-style.ipynb#

```
In [44]: 1 do_fit('2b', slice(1e-6,1e-4), pct_start=0.3)
```

Total time: 20:30

epoch	train_loss	valid_loss	pixel	feat_0	feat_1	feat_2	gram_0
1	1.633440	1.661697	0.338929	0.305611	0.264935	0.093316	0.240834
2	1.636672	1.662439	0.338239	0.305816	0.265915	0.093722	0.240812
3	1.630876	1.651365	0.336715	0.305728	0.264273	0.092895	0.236705
4	1.630698	1.652336	0.336751	0.305440	0.263674	0.092918	0.237603
5	1.628027	1.652598	0.335233	0.305314	0.264373	0.092932	0.238961
6	1.622456	1.652236	0.334720	0.305754	0.264049	0.092988	0.239227
7	1.614159	1.647799	0.333673	0.304889	0.264373	0.093023	0.237176
8	1.613826	1.650696	0.333413	0.305395	0.264024	0.092950	0.238165
9	1.616199	1.646712	0.332440	0.305456	0.264184	0.092992	0.237340
10	1.623145	1.649282	0.333545	0.305586	0.263895	0.092899	0.237928

Input / Prediction / Target







## **Test**

```
In [ ]:
             data_mr = (ImageImageList.from_folder(path_mr).random_split_by_pct(0.1, seed
                        .label from func(lambda x: path hr/x.name)
          2
          3
                        .transform(get_transforms(), size=(1280,1600), tfm_y=True)
          4
                        .databunch(bs=1).normalize(imagenet stats, do y=True))
          5
             data_mr.c = 3
In [ ]:
             learn.load('2b');
In [ ]:
             learn.data = data_mr
In [ ]:
             fn = data mr.valid ds.x.items[0]; fn
In [ ]:
             img = open image(fn); img.shape
             p,img_hr,b = learn.predict(img)
In [ ]:
             show image(img, figsize=(18,15), interpolation='nearest');
In [ ]:
In [ ]:
             Image(img_hr).show(figsize=(18,15))
In [ ]:
          1
```