

Flowers

相关论文

[resnet50](#)

[inceptionv3](#)

[xception](#)

[inceptionres](#)

数据集

数据集是在Kaggle上下载的[数据集](#).

训练集的相关信息都存储在json文件里,数据集相当大,训练集有70W张图片.

```
1 def get_data():
2     with open("test.json", 'r') as f:
3         test_dict = json.load(f)
4         test_images = test_dict["images"]
5     with open("train.json", 'r') as f:
6         train_dict = json.load(f)
7         annotations = train_dict["annotations"]
8
9     total_classes = 997 # test_dict['images']
10    total_test = len(test_images)
11    test_image_names = []
12
13    for index in range(total_test):
14        img_name = "test/"+test_images[index]["file_name"]
15        test_image_names.append(img_name)
16    test_image_names = np.array(test_image_names)
17    # print("-----train-----")
18    total_train = len(annotations)
19    train_image_names = []
20    train_image_labels = np.ones((total_train,))
21
22    for index in range(total_train):
23        img_name = "train/category_"+str(annotations[index]
24    ['category_id']+1)+"/"+str(annotations[index]["image_id"])+".jpg"
25        train_image_names.append(img_name)
26        train_image_labels[index] = annotations[index]["category_id"]
27    train_image_names = np.array(train_image_names)
28    # release memory
29    del annotations
30    del test_images
31    # train_test_split
32    train_x, valid_x, train_y, valid_y =
```

```

train_test_split(train_image_names,train_image_labels,test_size=0.2, random_state=42)
32
33     return train_x,valid_x,train_y,valid_y,test_image_names

```

预处理

计算了一下数据集的均值和标准差,打算用于训练(最后没有尝试),使用AWS P2.xlarge计算一代需要5个小时,时间太长,就没有计算到最后结果.

最后使用迁移学习,使用了各个模型在imagenet上的训练权值.

搭建模型

我使用了Resnet50,Xception,InceptionV3,InceptionRes四个模型进行迁移学习.

微调各个模型的输出层,避免对卷积层造成太大的更新,等输出层收敛了,在微调一下卷积层.

最后结合各个模型来导出特征向量,结合特征向量在搭建一个分类层,训练至收敛.

Resnet50

添加了dropout需要训练更多代才能收敛.

```

1 # ResModel
2 Input = KL.Input(shape=(224,224,3))
3 # pre-processing
4 process_input = KL.Lambda(res_preprocess_input)(Input)
5 resnet_50 = resnet50.ResNet50(include_top=False,
6     ..... weights='imagenet', input_tensor=process_input, pooling='avg')
7 # trainable
8 for layer in resnet_50.layers:
9     layer.trainable=False
10
11
12 # output_shape: batch_size * 2048
13 output = resnet_50.output
14 # dropout
15 # output = KL.Dropout(0.5)(output)
16 # classifier
17 output = KL.Dense(CLASS_COUNT,activation='softmax')(output) #
18     ,kernel_regularizer=regularizers.l2(0.001)
19 # model
20 res_model = KM.Model(resnet_50.input,output)
21
22 train_generator = MXGenerator((train_x,train_y),len(train_x),
23     ..... des_size=(224,224),means=None,stds=None,
24     ..... is_directory=True,batch_size=batch_size,shuffle=True,seed=0)
25
26 valid_generator = MXGenerator((valid_x,valid_y),len(valid_x),
27     ..... des_size=(224,224),means=None,stds=None,

```

```

27         is_directory=True, batch_size=batch_size, shuffle=True, seed=0)
28 adam = Adam(lr=0.001)
29 #
30 res_model.compile(optimizer=adam, loss='categorical_crossentropy', metrics=['accuracy'])
31
32 res_model.fit_generator(train_generator, len(train_x) // batch_size,
33                         epochs=1, verbose=1, validation_data=valid_generator,
34                         validation_steps=(len(valid_x) // batch_size))
35
36 # 放开所有层
37 for layer in resnet_50.layers:
38     layer.trainable=True
39
40 res_model.compile(optimizer=Adam(lr=0.001), loss='categorical_crossentropy', metrics=
41 ['accuracy'])
42 res_model.fit_generator(train_generator, len(train_x) // batch_size,
43                         epochs=epochs - 5, verbose=1, validation_data=valid_generator,
44                         validation_steps=(len(valid_x) // batch_size))
45
46 res_model.save_weights("res_model.h5")

```

其他几个模型也都是类似的处理,除了 inception_resnet_v2需要调节一下输出层以外,因为它的 avg pooling 之后的输出为 (batch, 1536),

conv_7b_bn (BatchNormalization)	(None, 5, 5, 1536)	4608	conv_7b[0][0]
conv_7b_ac (Activation)	(None, 5, 5, 1536)	0	conv_7b_bn[0][0]
global_average_pooling2d_4 (Glo	(None, 1536)	0	conv_7b_ac[0][0]

=====

Total params: 54,336,736
Trainable params: 54,276,192
Non-trainable params: 60,544

,所以添加一层全连接层让输出为 (batch, 2048) 与其他几个模型保持一致.

```

1 #output: (None, 1536)
2 output = inception_res.output
3 # dropout
4 output = KL.Dropout(0.5)(output)
5 # Dense
6 output = KL.Dense(2048, activation='relu')(output)
7 output = KL.Dropout(0.5)(output)
8 # classifier
9 output = KL.Dense(CLASS_COUNT, activation='softmax')(output)
10 # model
11 inc_res_model = KM.Model(inception_res.input, output, name='inc_res')

```

导出特征向量

通过几个微调后的模型分别进行特征向量的导出,其他模型也是相同的处理.

```

1 train_generator = MXGenerator((train_x, train_y), len(train_x),
2                               des_size=(299, 299), means=None, stds=None,
3                               is_directory=True, batch_size=batch_size, shuffle=True, seed=0)
4

```

```

5 valid_generator = MXGenerator((valid_x,valid_y),len(valid_x),
6                               des_size=(299,299),means=None,stds=None,
7                               is_directory=True,batch_size=batch_size,shuffle=True,seed=0)
8 # test_generator
9 test_generator = MXGenerator((test_images,None),len(test_images),
10                              des_size=(299,299),means=None,stds=None,
11                              is_directory=True,batch_size=batch_size,shuffle=False,seed=0)
12 #
13
14 xcep_feature_model = KM.Model(xcep_model.input,xcep_model.layers[-2].output)
15
16 xcep_train_feature_vector =
17     xcep_feature_model.predict_generator(train_generator,steps=len(train_x) // batch_size)
18 xcep_valid_feature_vector =
19     xcep_feature_model.predict_generator(valid_generator,steps=len(valid_x) // batch_size)
20 xcep_test_feature_vector = xcep_feature_model.predict_generator(test_generator,
21                                                                    steps=len(test_images) //
22                                                                    batch_size)
23
24 # save to h5
25 xception_h5 = h5py.File("xception.h5",'w')
26 xception_h5.create_dataset('train_x',
27                             xcep_train_feature_vector.shape,data=xcep_train_feature_vector)
28 xception_h5.create_dataset('train_label',
29                             trian_y.shape,data=trian_y)
30 xception_h5.create_dataset('valid_x',
31                             xcep_valid_feature_vector.shape,data=xcep_valid_feature_vector)
32 xception_h5.create_dataset('valid_label',
33                             valid_y.shape,data=valid_y)
34 xception_h5.create_dataset('test',
35                             xcep_test_feature_vector.shape,data=xcep_test_feature_vector)

```

结合特征向量以及构建新模型

分别导出了四个特征向量之后,就可以concatenate了,并训练与新模型.

```

1 X_train = []
2 y_train = []
3 X_valid = []
4 y_valid = []
5 test = []
6
7 for file_name in ['xception.h5','inc_res.h5','res_.h5','inc_.h5']:
8     with h5py.File(file_name,'r') as f:
9         X_train.append(f['train_x'])
10        X_valid.append(f['valid_x'])
11        y_train.append(f['y_train'])
12        y_valid.append(f['y_valid'])
13        test.append(f['test'])
14
15 X_train = np.concatenate(X_train, axis=1)
16 X_valid = np.concatenate(X_valid,axis=1)
17 test = np.concatenate(test,axis=1)
18 y_train = np.concatenate(y_train)

```

```

19 y_valid = np.concatenate(y_valid)
20
21 # build the network
22 merge_input = KL.Input(shape=X_train.shape[1:])
23 merge_dropout = KL.Dropout(0.5)(merge_input)
24 merge_classifier = KL.Dense(CLASS_COUNT,activation='softmax')(merge_dropout)
25 merge_model(inputs=[merge_input],outputs=[merge_classifier])
26
27 merge_model.compile(optimizer=Adam(lr=0.001),loss='categorical_crossentropy',metrics=
    ['accuracy'])
28
29 datagen = ImageGenerator()
30
31 merge_train_generator = datagen.flow((X_train,y_train))
32 merge_valid_generator = datagen.flow((X_valid,y_valid))
33
34 merge_model.fit_generator(train_generator,len(train_x) // batch_size,
35                           epochs=epochs, verbose=1,validation_data=valid_generator,
36                           validation_steps=(len(valid_x) // batch_size))

```

导出预测结果

使用融合模型进行预测最后结果,并生成predict.csv文件.

```

1 df = pd.read_csv("kaggle_sample_submission.csv")
2
3 images = np.zeros((224,224,3),dtype=np.int)
4
5 for i in range(len(test_images)):
6     img = cv2.imread(test_images[i])
7     img = cv2.cvtColor(img,cv2.BGRA2RGB)
8     img = cv2.resize(img,(224,224))
9     images[i] = img
10
11 # test_generator
12 test_generator = MXGenerator((test_images,None),len(test_images),
13                               des_size=(299,299),means=None,stds=None,
14                               is_directory=True,batch_size=batch_size,shuffle=False,seed=0)
15
16 y_pred = res_model.predict_generator(test_generator)
17
18 for i, fname in enumerate(test_generator filenames):
19     index = int(fname[fname.rfind('/')+1:fname.rfind('.')])
20     df.set_value(index-1, 'predict', y_pred[i])
21

```