fire-detection-jetson-save-ca

February 16, 2018

```
In [1]: from sklearn.datasets import load_files
        from keras.utils import np_utils
        import numpy as np
        from glob import glob
        import cv2
        import matplotlib.pyplot as plt
        #from keras.applications.resnet50 import preprocess_input, decode_predictions
        #from keras.applications.xception import Xception, preprocess_input
        #from keras.applications.inception_u3 import InceptionV3, preprocess_input
        #from keras.applications.ugg16 import VGG16, preprocess_input
        from keras.applications.vgg19 import VGG19, preprocess_input
Using TensorFlow backend.
0.0.1 Display nodes
```

```
In [2]: from __future__ import print_function
        def display_nodes(nodes):
            for i, node in enumerate(nodes):
                print('%d %s %s' % (i, node.name, node.op))
                [print(u' %d %s' % (i, n)) for i, n in enumerate(node.input)]
```

0.0.2 Work around GPU memory issues (not needed with GTX1080)

```
In [3]: NEED_GPU_MEM_WORKAROUND = False
        if (NEED_GPU_MEM_WORKAROUND):
            print('Working around TF GPU mem issues')
            import tensorflow as tf
            import keras.backend.tensorflow_backend as ktf
            def get_session(gpu_fraction=0.6):
                gpu_options = tf.GPUOptions(per_process_gpu_memory_fraction=gpu_fraction,
                                        allow_growth=True)
                return tf.Session(config=tf.ConfigProto(gpu_options=gpu_options))
            ktf.set_session(get_session())
```

```
In [4]: img_width, img_height = 224, 224 # change based on the shape/structure of your images
        num_classes = 2 # Fire or Safe
        # define function to load train, test, and validation datasets
        def load_dataset(path):
            data = load_files(path)
            fire_files = np.array(data['filenames'])
            fire_targets = np_utils.to_categorical(np.array(data['target']), num_classes)
            return fire_files, fire_targets
        # load train, test, and validation datasets
        train_files, train_targets = load_dataset('fireImages/train')
        valid_files, valid_targets = load_dataset('fireImages/valid')
        test_files, test_targets = load_dataset('fireImages/test')
        # load list of fire classes
        class_names = [item[21:-1] for item in sorted(glob("fireImages/train/*/"))]
        # print statistics about the dataset
        print('There are %d total fire categories.' % len(class_names))
       print('There are %s total fire images.\n' % len(np.hstack([train_files, valid_files, tes
        print('There are %d training fire images.' % len(train_files))
        print('There are %d validation fire images.' % len(valid_files))
        print('There are %d test fire images.'% len(test_files))
There are 2 total fire categories.
There are 1069 total fire images.
There are 1015 training fire images.
There are 45 validation fire images.
There are 9 test fire images.
In [5]: class_names
Out[5]: ['Fire', 'Safe']
In [6]: from keras.preprocessing import image
        from tqdm import tqdm
        def path_to_tensor(img_path):
            # loads RGB image as PIL.Image.Image type
            img = image.load_img(img_path, target_size=(img_width, img_height))
            # convert PIL. Image. Image type to 3D tensor with shape (224, 224, 3)
            x = image.img_to_array(img)
            # convert 3D tensor to 4D tensor with shape (1, 224, 224, 3) and return 4D tensor
            return np.expand_dims(x, axis=0)
        def paths_to_tensor(img_paths):
```

```
list_of_tensors = [path_to_tensor(img_path) for img_path in tqdm(img_paths)]
    return np.vstack(list_of_tensors)

In [7]: from PIL import ImageFile
    from PIL import Image
    ImageFile.LOAD_TRUNCATED_IMAGES = True

# pre-process the data for Keras
    train_tensors = preprocess_input( paths_to_tensor(train_files) )
    valid_tensors = preprocess_input( paths_to_tensor(valid_files) )
    test_tensors = preprocess_input( paths_to_tensor(test_files) )

100%|| 1015/1015 [00:30<00:00, 33.17it/s]
100%|| 45/45 [00:01<00:00, 31.89it/s]
100%|| 9/9 [00:00<00:00, 55.78it/s]</pre>

0.0.3 Detect fire with the most accurate model

In [8]: ### Define the architecture.
```

```
In [8]: ### Define the architecture.
    from keras.layers import Conv2D, MaxPooling2D, GlobalAveragePooling2D
    from keras.layers import Dropout, Flatten, Dense
    from keras.models import Sequential
    from keras import applications

from keras.applications.resnet50 import preprocess_input, decode_predictions
VGG16_model = applications.VGG16(input_shape=(img_width, img_height, 3), weights =
InceptionV3_model = applications.InceptionV3(input_shape=(img_width, img_height, 3), weight
Xception_model = applications.Xception(input_shape=(img_width, img_height, 3), weight
ResNet50_model = applications.ResNet50(input_shape=(img_width, img_height, 3), weight
VGG19_model = applications.VGG19(input_shape=(img_width, img_height, 3), weights =
```

0.0.4 Pick the most accurate model for this application

block1_conv2 (Conv2D)	(None, 224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None, 112, 112, 64)	0
block2_conv1 (Conv2D)	(None, 112, 112, 128)	73856
block2_conv2 (Conv2D)	(None, 112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None, 56, 56, 128)	0
block3_conv1 (Conv2D)	(None, 56, 56, 256)	295168
block3_conv2 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv3 (Conv2D)	(None, 56, 56, 256)	590080
block3_conv4 (Conv2D)	(None, 56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None, 28, 28, 256)	0
block4_conv1 (Conv2D)	(None, 28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None, 28, 28, 512)	2359808
block4_conv4 (Conv2D)	(None, 28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None, 14, 14, 512)	0
block5_conv1 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None, 14, 14, 512)	2359808
block5_conv4 (Conv2D)	(None, 14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None, 7, 7, 512)	0
Total params: 20,024,384		_

Total params: 20,024,384 Trainable params: 20,024,384 Non-trainable params: 0

0.0.5 Transfer learning (freeze base model layers)

```
In [11]: for i, layer in enumerate(base_model.layers):
                               print(i, layer.name)
0 input_5
1 block1_conv1
2 block1_conv2
3 block1_pool
4 block2_conv1
5 block2_conv2
6 block2_pool
7 block3_conv1
8 block3 conv2
9 block3_conv3
10 block3_conv4
11 block3_pool
12 block4_conv1
13 block4_conv2
14 block4_conv3
15 block4_conv4
16 block4_pool
17 block5_conv1
18 block5_conv2
19 block5_conv3
20 block5_conv4
21 block5_pool
In [12]: # Freeze the layers which you don't want to train.
                       for layer in base_model.layers:
                                  layer.trainable = False
                        # Here I am freezing the first 5 layers.
                        # for layer in model.layers[:5]:
                                 layer.trainable = False
In [13]: # add a global spatial average pooling layer
                       x = base_model.output
                       x = GlobalAveragePooling2D()(x)
                       \# x = Dropout(0.45)(x)
                       # and a logistic layer we have num_classes classes
                       predictions = Dense(num_classes, activation='softmax')(x)
In [14]: from keras.models import Model
                        # this is the model we will train
                      model = Model(inputs=base_model.input, outputs=predictions)
In [15]: # compile the model (should be done *after* setting layers to non-trainable)
                       model.compile(optimizer='rmsprop', loss='categorical_crossentropy', metrics=['accuracy'
                       \#model.compile(loss='categorical\_crossentropy', optimizer='adadelta', metrics=['accuracy', optimizer='adade', opt
```

In [16]: model.summary()

Layer (type)	Output	Shape	 Param #
input_5 (InputLayer)	(None,	224, 224, 3)	0
block1_conv1 (Conv2D)	(None,	224, 224, 64)	1792
block1_conv2 (Conv2D)	(None,	224, 224, 64)	36928
block1_pool (MaxPooling2D)	(None,	112, 112, 64)	0
block2_conv1 (Conv2D)	(None,	112, 112, 128)	73856
block2_conv2 (Conv2D)	(None,	112, 112, 128)	147584
block2_pool (MaxPooling2D)	(None,	56, 56, 128)	0
block3_conv1 (Conv2D)	(None,	56, 56, 256)	295168
block3_conv2 (Conv2D)	(None,	56, 56, 256)	590080
block3_conv3 (Conv2D)	(None,	56, 56, 256)	590080
block3_conv4 (Conv2D)	(None,	56, 56, 256)	590080
block3_pool (MaxPooling2D)	(None,	28, 28, 256)	0
block4_conv1 (Conv2D)	(None,	28, 28, 512)	1180160
block4_conv2 (Conv2D)	(None,	28, 28, 512)	2359808
block4_conv3 (Conv2D)	(None,	28, 28, 512)	2359808
block4_conv4 (Conv2D)	(None,	28, 28, 512)	2359808
block4_pool (MaxPooling2D)	(None,	14, 14, 512)	0
block5_conv1 (Conv2D)	(None,	14, 14, 512)	2359808
block5_conv2 (Conv2D)	(None,	14, 14, 512)	2359808
block5_conv3 (Conv2D)	(None,	14, 14, 512)	2359808
block5_conv4 (Conv2D)	(None,	14, 14, 512)	2359808
block5_pool (MaxPooling2D)	(None,	7, 7, 512)	0

```
(None, 2)
dense_1 (Dense)
                                                    1026
______
Total params: 20,025,410
Trainable params: 1,026
Non-trainable params: 20,024,384
  Note that we only train 1,539 params and the rest are frozen:
  • Total params: 20,025,410
  • Trainable params: 1,026
  • Non-trainable params: 20,024,384
In [17]: from keras.callbacks import ModelCheckpoint
        from keras.callbacks import TensorBoard
        tbCallback = TensorBoard(log_dir='./Graph', histogram_freq=0, write_graph=True, write_i
        import keras.backend.tensorflow_backend as K
        # train the model
        checkpointer = ModelCheckpoint(filepath='firemodel.weights.best.hdf5', verbose=3, save_
        hist = model.fit(train_tensors, train_targets, batch_size=64, epochs=10,
              validation_data=(valid_tensors, valid_targets), callbacks=[checkpointer, tbCallbacks=
              verbose=2) #, shuffle=True)
Train on 1015 samples, validate on 45 samples
Epoch 1/10
Epoch 00001: val_loss improved from inf to 0.70525, saving model to firemodel.weights.best.hdf5
 - 12s - loss: 1.2621 - acc: 0.6473 - val_loss: 0.7052 - val_acc: 0.7556
Epoch 2/10
Epoch 00002: val_loss improved from 0.70525 to 0.44532, saving model to firemodel.weights.best.h
 - 6s - loss: 0.5690 - acc: 0.8158 - val_loss: 0.4453 - val_acc: 0.8000
Epoch 3/10
Epoch 00003: val_loss improved from 0.44532 to 0.25189, saving model to firemodel.weights.best.h
- 6s - loss: 0.3952 - acc: 0.8700 - val_loss: 0.2519 - val_acc: 0.8889
Epoch 00004: val_loss improved from 0.25189 to 0.18459, saving model to firemodel.weights.best.h
 - 6s - loss: 0.3114 - acc: 0.9025 - val_loss: 0.1846 - val_acc: 0.9111
```

global_average_pooling2d_1 ((None, 512)

- 6s - loss: 0.2497 - acc: 0.9153 - val_loss: 0.1339 - val_acc: 0.9111

- 6s - loss: 0.2169 - acc: 0.9320 - val_loss: 0.1007 - val_acc: 0.9333

Epoch 6/10

Epoch 00005: val_loss improved from 0.18459 to 0.13390, saving model to firemodel.weights.best.h

Epoch 00006: val_loss improved from 0.13390 to 0.10070, saving model to firemodel.weights.best.h

```
Epoch 7/10
Epoch 00007: val_loss did not improve
 - 6s - loss: 0.1907 - acc: 0.9360 - val_loss: 0.1242 - val_acc: 0.9556
Epoch 8/10
Epoch 00008: val_loss did not improve
 - 6s - loss: 0.1607 - acc: 0.9419 - val_loss: 0.1383 - val_acc: 0.9556
Epoch 00009: val_loss improved from 0.10070 to 0.09260, saving model to firemodel.weights.best.h
- 6s - loss: 0.1511 - acc: 0.9409 - val_loss: 0.0926 - val_acc: 0.9556
Epoch 10/10
Epoch 00010: val_loss improved from 0.09260 to 0.06809, saving model to firemodel.weights.best.h
 - 6s - loss: 0.1236 - acc: 0.9547 - val_loss: 0.0681 - val_acc: 0.9778
In [18]: for i, layer in enumerate(model.layers):
            print(i, layer.name)
0 input_5
1 block1_conv1
2 block1_conv2
3 block1_pool
4 block2_conv1
5 block2_conv2
6 block2_pool
7 block3_conv1
8 block3_conv2
9 block3_conv3
10 block3_conv4
11 block3_pool
12 block4_conv1
13 block4_conv2
14 block4_conv3
15 block4_conv4
16 block4_pool
17 block5_conv1
18 block5_conv2
19 block5_conv3
20 block5_conv4
21 block5_pool
22 global_average_pooling2d_1
23 dense_1
In [19]: import cv2
         import matplotlib.pyplot as plt
         # summarize history for accuracy
         plt.figure(figsize=(4,4), dpi=100 )
```

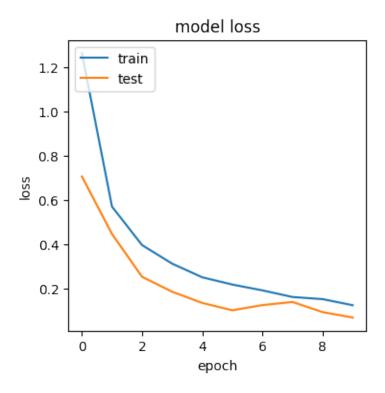
```
plt.plot(hist.history['acc'])
plt.plot(hist.history['val_acc'])
plt.title('model accuracy')
plt.ylabel('accuracy')
plt.xlabel('epoch')
plt.legend(['train', 'test'], loc='upper left')
plt.show()
plt.savefig('training1.png', dpi=300)
```

model accuracy train 0.95 test 0.90 0.85 accuracy 0.80 0.75 0.70 0.65 0 2 4 6 8

epoch

<matplotlib.figure.Figure at 0x7eff2a5ff190>

```
In [20]: # summarize history for loss
    plt.figure(figsize=(4,4), dpi=100)
    plt.plot(hist.history['loss'])
    plt.plot(hist.history['val_loss'])
    plt.title('model loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.legend(['train', 'test'], loc='upper left')
    plt.show()
```



0.0.6 Load the Model with the Best Validation Loss

0.1 Test the Fire detection Model

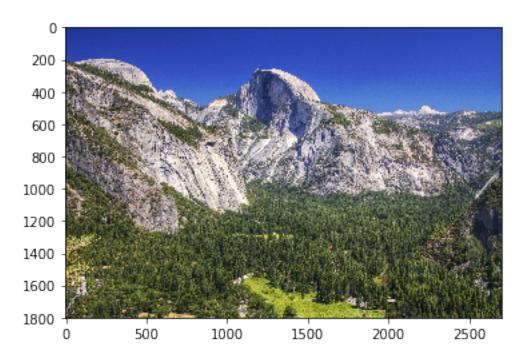
Validation accuracy: 97.0000%

0.1.1 Calculate classification accuracy on the test dataset.

```
In [25]: def detect_fire(img_path):
            predicted_vector = model.predict(preprocess_input(path_to_tensor(img_path)))
            return class_names[np.argmax(predicted_vector)]
In [26]: import glob
        from PIL import Image
        from IPython import display
        path = "test_images/*"
        for fname in glob.glob(path):
            fire_detection = detect_fire( fname )
            display(plt.gcf())
            print('_____
            print('')
            if (fire_detection=='Fire'):
                print ('\033[1m'+'\033[91m'+'ALARM: Detected '+fire_detection.replace("_", ""))
            elif (fire_detection=='Smoke'):
                print ('\033[1m'+'\033[91m'+'ALARM: Detected '+fire_detection.replace("_", ""))
            else:
                print ('\033[1m'+'\033[92m'+'Looking good: ' +fire_detection.replace("_", ""))
            plt.imshow(Image.open(fname))
             #raw_input()
```

<matplotlib.figure.Figure at 0x7eff1ca783d0>

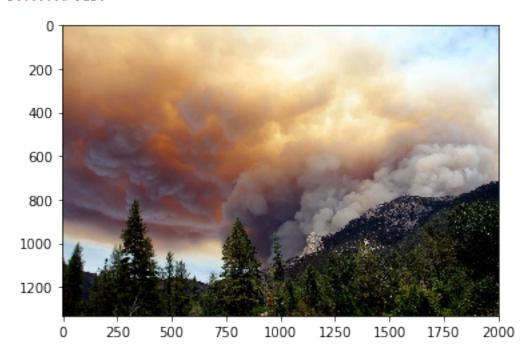
Looking good: Safe



ALARM: Detected Fire



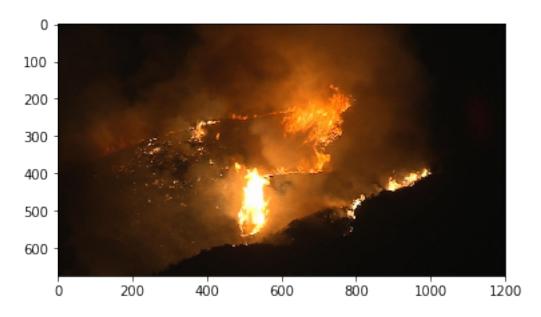
ALARM: Detected Fire



Looking good: Safe



ALARM: Detected Fire



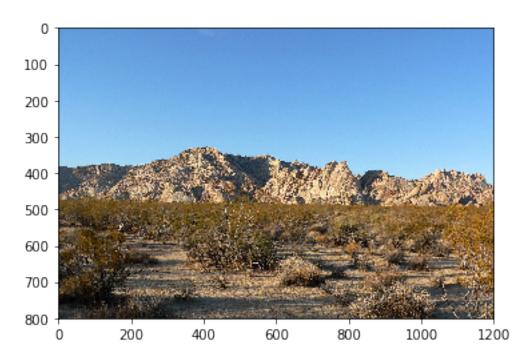
ALARM: Detected Fire



Looking good: Safe



Looking good: Safe



ALARM: Detected Fire



1 Convert the Keras / TF model to something that Jetson TX2 understands

1.0.1 Freeze the graph

```
INFO:tensorflow:Restoring parameters from fire_detector_saved_ckpt-0
INFO:tensorflow:Froze 34 variables.
Converted 34 variables to const ops.
127 ops in the final graph.
In [33]: # read frozen graph and display nodes
        graph = tf.GraphDef()
        with tf.gfile.Open('frozen_fire_detector.pb', 'r') as f:
             data = f.read()
             graph.ParseFromString(data)
         display_nodes(graph.node)
0 input_5 Placeholder
1 block1_conv1_2/kernel Const
2 block1_conv1_2/kernel/read Identity
0 block1_conv1_2/kernel
3 block1_conv1_2/bias Const
4 block1_conv1_2/bias/read Identity
0 block1_conv1_2/bias
5 block1_conv1_2/convolution Conv2D
0 input_5
1 block1_conv1_2/kernel/read
6 block1_conv1_2/BiasAdd BiasAdd
0 block1_conv1_2/convolution
1 block1_conv1_2/bias/read
7 block1_conv1_2/Relu Relu
0 block1_conv1_2/BiasAdd
8 block1 conv2 2/kernel Const
9 block1_conv2_2/kernel/read Identity
0 block1 conv2 2/kernel
10 block1_conv2_2/bias Const
11 block1_conv2_2/bias/read Identity
0 block1_conv2_2/bias
12 block1_conv2_2/convolution Conv2D
0 block1_conv1_2/Relu
1 block1_conv2_2/kernel/read
13 block1_conv2_2/BiasAdd BiasAdd
0 block1_conv2_2/convolution
 1 block1_conv2_2/bias/read
14 block1_conv2_2/Relu Relu
0 block1_conv2_2/BiasAdd
15 block1_pool_1/MaxPool MaxPool
0 block1 conv2 2/Relu
16 block2_conv1_1/kernel Const
17 block2_conv1_1/kernel/read Identity
0 block2_conv1_1/kernel
```

- 18 block2_conv1_1/bias Const
- 19 block2_conv1_1/bias/read Identity
- 0 block2_conv1_1/bias
- 20 block2_conv1_1/convolution Conv2D
- 0 block1_pool_1/MaxPool
- 1 block2_conv1_1/kernel/read
- 21 block2_conv1_1/BiasAdd BiasAdd
- 0 block2_conv1_1/convolution
- 1 block2_conv1_1/bias/read
- 22 block2_conv1_1/Relu Relu
- 0 block2_conv1_1/BiasAdd
- 23 block2_conv2_1/kernel Const
- 24 block2_conv2_1/kernel/read Identity
- 0 block2_conv2_1/kernel
- 25 block2_conv2_1/bias Const
- 26 block2_conv2_1/bias/read Identity
- 0 block2_conv2_1/bias
- 27 block2_conv2_1/convolution Conv2D
- 0 block2_conv1_1/Relu
- 1 block2_conv2_1/kernel/read
- 28 block2_conv2_1/BiasAdd BiasAdd
- 0 block2_conv2_1/convolution
- 1 block2_conv2_1/bias/read
- 29 block2_conv2_1/Relu Relu
- 0 block2_conv2_1/BiasAdd
- 30 block2_pool_2/MaxPool MaxPool
- 0 block2_conv2_1/Relu
- 31 block3_conv1_1/kernel Const
- 32 block3_conv1_1/kernel/read Identity
- 0 block3_conv1_1/kernel
- 33 block3_conv1_1/bias Const
- 34 block3_conv1_1/bias/read Identity
- 0 block3_conv1_1/bias
- 35 block3_conv1_1/convolution Conv2D
- 0 block2_pool_2/MaxPool
- 1 block3_conv1_1/kernel/read
- 36 block3_conv1_1/BiasAdd BiasAdd
- 0 block3_conv1_1/convolution
- 1 block3_conv1_1/bias/read
- 37 block3_conv1_1/Relu Relu
- 0 block3_conv1_1/BiasAdd
- 38 block3_conv2_1/kernel Const
- 39 block3_conv2_1/kernel/read Identity
- 0 block3_conv2_1/kernel
- 40 block3_conv2_1/bias Const
- 41 block3_conv2_1/bias/read Identity
- 0 block3_conv2_1/bias
- 42 block3_conv2_1/convolution Conv2D

- 0 block3_conv1_1/Relu
- 1 block3_conv2_1/kernel/read
- 43 block3_conv2_1/BiasAdd BiasAdd
- 0 block3_conv2_1/convolution
- 1 block3_conv2_1/bias/read
- 44 block3_conv2_1/Relu Relu
- 0 block3_conv2_1/BiasAdd
- 45 block3_conv3_1/kernel Const
- 46 block3_conv3_1/kernel/read Identity
- 0 block3_conv3_1/kernel
- 47 block3_conv3_1/bias Const
- 48 block3_conv3_1/bias/read Identity
- 0 block3_conv3_1/bias
- 49 block3 conv3 1/convolution Conv2D
- 0 block3_conv2_1/Relu
- 1 block3_conv3_1/kernel/read
- 50 block3_conv3_1/BiasAdd BiasAdd
- 0 block3_conv3_1/convolution
- 1 block3_conv3_1/bias/read
- 51 block3_conv3_1/Relu Relu
- 0 block3_conv3_1/BiasAdd
- 52 block3_conv4/kernel Const
- 53 block3_conv4/kernel/read Identity
- 0 block3_conv4/kernel
- 54 block3_conv4/bias Const
- 55 block3_conv4/bias/read Identity
- 0 block3_conv4/bias
- 56 block3_conv4/convolution Conv2D
- 0 block3 conv3 1/Relu
- 1 block3_conv4/kernel/read
- 57 block3 conv4/BiasAdd BiasAdd
- 0 block3_conv4/convolution
- 1 block3_conv4/bias/read
- 58 block3_conv4/Relu Relu
- 0 block3_conv4/BiasAdd
- 59 block3_pool_2/MaxPool MaxPool
- 0 block3_conv4/Relu
- 60 block4_conv1_1/kernel Const
- 61 block4_conv1_1/kernel/read Identity
- 0 block4_conv1_1/kernel
- 62 block4_conv1_1/bias Const
- 63 block4_conv1_1/bias/read Identity
- 0 block4_conv1_1/bias
- 64 block4_conv1_1/convolution Conv2D
- 0 block3_pool_2/MaxPool
- 1 block4_conv1_1/kernel/read
- 65 block4_conv1_1/BiasAdd BiasAdd
- 0 block4_conv1_1/convolution

- 1 block4_conv1_1/bias/read
- 66 block4_conv1_1/Relu Relu
- 0 block4_conv1_1/BiasAdd
- 67 block4_conv2_1/kernel Const
- 68 block4_conv2_1/kernel/read Identity
- 0 block4_conv2_1/kernel
- 69 block4_conv2_1/bias Const
- 70 block4_conv2_1/bias/read Identity
- 0 block4_conv2_1/bias
- 71 block4_conv2_1/convolution Conv2D
- 0 block4_conv1_1/Relu
- 1 block4_conv2_1/kernel/read
- 72 block4_conv2_1/BiasAdd BiasAdd
- 0 block4_conv2_1/convolution
- 1 block4_conv2_1/bias/read
- 73 block4_conv2_1/Relu Relu
- 0 block4_conv2_1/BiasAdd
- 74 block4_conv3_1/kernel Const
- 75 block4_conv3_1/kernel/read Identity
- 0 block4_conv3_1/kernel
- 76 block4_conv3_1/bias Const
- 77 block4_conv3_1/bias/read Identity
- 0 block4_conv3_1/bias
- 78 block4_conv3_1/convolution Conv2D
- 0 block4_conv2_1/Relu
- 1 block4_conv3_1/kernel/read
- 79 block4_conv3_1/BiasAdd BiasAdd
- 0 block4_conv3_1/convolution
- 1 block4 conv3 1/bias/read
- 80 block4_conv3_1/Relu Relu
- O block4 conv3 1/BiasAdd
- 81 block4_conv4/kernel Const
- 82 block4_conv4/kernel/read Identity
- 0 block4_conv4/kernel
- 83 block4_conv4/bias Const
- 84 block4_conv4/bias/read Identity
- 0 block4_conv4/bias
- 85 block4_conv4/convolution Conv2D
- 0 block4_conv3_1/Relu
- 1 block4_conv4/kernel/read
- 86 block4_conv4/BiasAdd BiasAdd
- 0 block4_conv4/convolution
- 1 block4_conv4/bias/read
- 87 block4 conv4/Relu Relu
- 0 block4_conv4/BiasAdd
- 88 block4_pool_2/MaxPool MaxPool
- 0 block4_conv4/Relu
- 89 block5_conv1_1/kernel Const

- 90 block5_conv1_1/kernel/read Identity
- 0 block5_conv1_1/kernel
- 91 block5_conv1_1/bias Const
- 92 block5_conv1_1/bias/read Identity
- 0 block5_conv1_1/bias
- 93 block5_conv1_1/convolution Conv2D
- 0 block4_pool_2/MaxPool
- 1 block5_conv1_1/kernel/read
- 94 block5_conv1_1/BiasAdd BiasAdd
- 0 block5_conv1_1/convolution
- 1 block5_conv1_1/bias/read
- 95 block5_conv1_1/Relu Relu
- 0 block5_conv1_1/BiasAdd
- 96 block5_conv2_1/kernel Const
- 97 block5_conv2_1/kernel/read Identity
- 0 block5_conv2_1/kernel
- 98 block5_conv2_1/bias Const
- 99 block5_conv2_1/bias/read Identity
- 0 block5_conv2_1/bias
- 100 block5_conv2_1/convolution Conv2D
- 0 block5_conv1_1/Relu
- 1 block5_conv2_1/kernel/read
- 101 block5_conv2_1/BiasAdd BiasAdd
- 0 block5_conv2_1/convolution
- 1 block5_conv2_1/bias/read
- 102 block5_conv2_1/Relu Relu
- O block5_conv2_1/BiasAdd
- 103 block5_conv3_1/kernel Const
- 104 block5_conv3_1/kernel/read Identity
- 0 block5_conv3_1/kernel
- 105 block5_conv3_1/bias Const
- 106 block5_conv3_1/bias/read Identity
- 0 block5_conv3_1/bias
- 107 block5_conv3_1/convolution Conv2D
- 0 block5_conv2_1/Relu
- 1 block5_conv3_1/kernel/read
- 108 block5_conv3_1/BiasAdd BiasAdd
- 0 block5_conv3_1/convolution
- 1 block5_conv3_1/bias/read
- 109 block5_conv3_1/Relu Relu
- 0 block5_conv3_1/BiasAdd
- 110 block5_conv4/kernel Const
- 111 block5_conv4/kernel/read Identity
- 0 block5 conv4/kernel
- 112 block5 conv4/bias Const
- 113 block5_conv4/bias/read Identity
- 0 block5_conv4/bias
- 114 block5_conv4/convolution Conv2D

```
0 block5_conv3_1/Relu
```

- 1 block5_conv4/kernel/read
- 115 block5_conv4/BiasAdd BiasAdd
- 0 block5_conv4/convolution
- 1 block5_conv4/bias/read
- 116 block5_conv4/Relu Relu
- 0 block5_conv4/BiasAdd
- 117 block5_pool_1/MaxPool MaxPool
- 0 block5_conv4/Relu
- 118 global_average_pooling2d_1/Mean/reduction_indices Const
- 119 global_average_pooling2d_1/Mean Mean
- 0 block5_pool_1/MaxPool
- 1 global_average_pooling2d_1/Mean/reduction_indices
- 120 dense_1/kernel Const
- 121 dense_1/kernel/read Identity
- 0 dense 1/kernel
- 122 dense_1/bias Const
- 123 dense_1/bias/read Identity
- 0 dense_1/bias
- 124 dense_1/MatMul MatMul
- O global_average_pooling2d_1/Mean
- 1 dense_1/kernel/read
- 125 dense 1/BiasAdd BiasAdd
- 0 dense_1/MatMul
- 1 dense_1/bias/read
- 126 dense_1/Softmax Softmax
- O dense_1/BiasAdd

1.0.2 Convert to UFF

```
In [34]: !convert-to-uff tensorflow --input-file frozen_fire_detector.pb -l
Loading frozen_fire_detector.pb
1 Placeholder: "input_5"
2 Const: "block1_conv1_2/kernel"
3 Identity: "block1_conv1_2/kernel/read"
4 Const: "block1_conv1_2/bias"
5 Identity: "block1_conv1_2/bias/read"
```

6 Conv2D: "block1_conv1_2/convolution"
7 BiasAdd: "block1_conv1_2/BiasAdd"

8 Relu: "block1_conv1_2/Relu"

9 Const: "block1_conv2_2/kernel"

10 Identity: "block1_conv2_2/kernel/read"

11 Const: "block1_conv2_2/bias"

12 Identity: "block1_conv2_2/bias/read"

13 Conv2D: "block1_conv2_2/convolution"

14 BiasAdd: "block1_conv2_2/BiasAdd"

- 15 Relu: "block1_conv2_2/Relu"
- 16 MaxPool: "block1_pool_1/MaxPool"
- 17 Const: "block2_conv1_1/kernel"
- 18 Identity: "block2_conv1_1/kernel/read"
- 19 Const: "block2_conv1_1/bias"
- 20 Identity: "block2_conv1_1/bias/read"
- 21 Conv2D: "block2_conv1_1/convolution"
- 22 BiasAdd: "block2_conv1_1/BiasAdd"
- 23 Relu: "block2_conv1_1/Relu"
- 24 Const: "block2_conv2_1/kernel"
- 25 Identity: "block2_conv2_1/kernel/read"
- 26 Const: "block2_conv2_1/bias"
- 27 Identity: "block2_conv2_1/bias/read"
- 28 Conv2D: "block2_conv2_1/convolution"
- 29 BiasAdd: "block2_conv2_1/BiasAdd"
- 30 Relu: "block2_conv2_1/Relu"
- 31 MaxPool: "block2_pool_2/MaxPool"
- 32 Const: "block3_conv1_1/kernel"
- 33 Identity: "block3_conv1_1/kernel/read"
- 34 Const: "block3_conv1_1/bias"
- 35 Identity: "block3_conv1_1/bias/read"
- 36 Conv2D: "block3_conv1_1/convolution"
- 37 BiasAdd: "block3_conv1_1/BiasAdd"
- 38 Relu: "block3_conv1_1/Relu"
- 39 Const: "block3_conv2_1/kernel"
- 40 Identity: "block3_conv2_1/kernel/read"
- 41 Const: "block3_conv2_1/bias"
- 42 Identity: "block3_conv2_1/bias/read"
- 43 Conv2D: "block3 conv2 1/convolution"
- 44 BiasAdd: "block3_conv2_1/BiasAdd"
- 45 Relu: "block3 conv2 1/Relu"
- 46 Const: "block3_conv3_1/kernel"
- 47 Identity: "block3_conv3_1/kernel/read"
- 48 Const: "block3_conv3_1/bias"
- 49 Identity: "block3_conv3_1/bias/read"
- 50 Conv2D: "block3_conv3_1/convolution"
- 51 BiasAdd: "block3_conv3_1/BiasAdd"
- 52 Relu: "block3_conv3_1/Relu"
- 53 Const: "block3_conv4/kernel"
- 54 Identity: "block3_conv4/kernel/read"
- 55 Const: "block3_conv4/bias"
- 56 Identity: "block3_conv4/bias/read"
- 57 Conv2D: "block3_conv4/convolution"
- 58 BiasAdd: "block3 conv4/BiasAdd"
- 59 Relu: "block3_conv4/Relu"
- 60 MaxPool: "block3_pool_2/MaxPool"
- 61 Const: "block4_conv1_1/kernel"
- 62 Identity: "block4_conv1_1/kernel/read"

- 63 Const: "block4_conv1_1/bias"
- 64 Identity: "block4_conv1_1/bias/read"
- 65 Conv2D: "block4_conv1_1/convolution"
- 66 BiasAdd: "block4_conv1_1/BiasAdd"
- 67 Relu: "block4_conv1_1/Relu"
- 68 Const: "block4_conv2_1/kernel"
- 69 Identity: "block4_conv2_1/kernel/read"
- 70 Const: "block4_conv2_1/bias"
- 71 Identity: "block4_conv2_1/bias/read"
- 72 Conv2D: "block4_conv2_1/convolution"
- 73 BiasAdd: "block4_conv2_1/BiasAdd"
- 74 Relu: "block4_conv2_1/Relu"
- 75 Const: "block4_conv3_1/kernel"
- 76 Identity: "block4_conv3_1/kernel/read"
- 77 Const: "block4_conv3_1/bias"
- 78 Identity: "block4_conv3_1/bias/read"
- 79 Conv2D: "block4_conv3_1/convolution"
- 80 BiasAdd: "block4_conv3_1/BiasAdd"
- 81 Relu: "block4_conv3_1/Relu"
- 82 Const: "block4_conv4/kernel"
- 83 Identity: "block4_conv4/kernel/read"
- 84 Const: "block4_conv4/bias"
- 85 Identity: "block4_conv4/bias/read"
- 86 Conv2D: "block4_conv4/convolution"
- 87 BiasAdd: "block4_conv4/BiasAdd"
- 88 Relu: "block4_conv4/Relu"
- 89 MaxPool: "block4_pool_2/MaxPool"
- 90 Const: "block5_conv1_1/kernel"
- 91 Identity: "block5 conv1 1/kernel/read"
- 92 Const: "block5_conv1_1/bias"
- 93 Identity: "block5_conv1_1/bias/read"
- 94 Conv2D: "block5_conv1_1/convolution"
- 95 BiasAdd: "block5_conv1_1/BiasAdd"
- 96 Relu: "block5_conv1_1/Relu"
- 97 Const: "block5_conv2_1/kernel"
- 98 Identity: "block5_conv2_1/kernel/read"
- 99 Const: "block5_conv2_1/bias"
- 100 Identity: "block5_conv2_1/bias/read"
- 101 Conv2D: "block5_conv2_1/convolution"
- 102 BiasAdd: "block5_conv2_1/BiasAdd"
- 103 Relu: "block5_conv2_1/Relu"
- 104 Const: "block5_conv3_1/kernel"
- 105 Identity: "block5_conv3_1/kernel/read"
- 106 Const: "block5 conv3 1/bias"
- 107 Identity: "block5_conv3_1/bias/read"
- 108 Conv2D: "block5_conv3_1/convolution"
- 109 BiasAdd: "block5_conv3_1/BiasAdd"
- 110 Relu: "block5_conv3_1/Relu"

```
111 Const: "block5_conv4/kernel"
112 Identity: "block5_conv4/kernel/read"
113 Const: "block5_conv4/bias"
114 Identity: "block5_conv4/bias/read"
115 Conv2D: "block5_conv4/convolution"
116 BiasAdd: "block5_conv4/BiasAdd"
117 Relu: "block5_conv4/Relu"
118 MaxPool: "block5_pool_1/MaxPool"
119 Const: "global_average_pooling2d_1/Mean/reduction_indices"
120 Mean: "global_average_pooling2d_1/Mean"
121 Const: "dense_1/kernel"
122 Identity: "dense_1/kernel/read"
123 Const: "dense_1/bias"
124 Identity: "dense_1/bias/read"
125 MatMul: "dense_1/MatMul"
126 BiasAdd: "dense_1/BiasAdd"
127 Softmax: "dense_1/Softmax"
```

1.0.3 Convert to uff

```
In [35]: !convert-to-uff tensorflow -o fire_detector.uff --input-file frozen_fire_detector.pb -C
Loading frozen_fire_detector.pb
Using output node dense_1/Softmax
Converting to UFF graph
No. nodes: 93
UFF Output written to fire_detector.uff

In [36]: !ls -lh fire_detector.uff
-rw-rw-r-- 1 ngeorgis ngeorgis 77M Feb 16 14:55 fire_detector.uff
```

1.0.4 Visualize model using tensorboard

