import os import h5py import numpy as np import json import urllib.request import numpy as np import matplotlib.pyplot as plt import pandas as pd import seaborn as sns from IPython.display import Image, display, clear_output from sklearn.metrics import classification report, confusion matrix

%matplotlib inline sns.set style('whitegrid')

In [2]:

from keras import optimizers from keras.applications.vgg16 import VGG16 from keras.models import Sequential, load_model, Model from keras.layers import Conv2D, MaxPooling2D, ZeroPadding2D, Activation, Dropout, Flatten, Dense, Input from keras.regularizers import I2, I1 from keras.utils.np_utils import to categorical from keras.preprocessing.image import ImageDataGenerator, array to img, img to array, load_img from keras.callbacks import ModelCheckpoint, History from keras import backend as K from keras.utils.data utils import get file

Using TensorFlow backend.

In [3]:

```
def plot metrics(hist, stop=50):
  fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(10,4))
  axes = axes.flatten()
  axes[0].plot(range(stop), hist['acc'], label='Training', color='#FF533D')
  axes[0].plot(range(stop), hist['val acc'], label='Validation', color='#03507E')
  axes[0].set_title('Accuracy')
  axes[0].set_ylabel('Accuracy')
  axes[0].set xlabel('Epoch')
  axes[0].legend(loc='lower right')
  axes[1].plot(range(stop), hist['loss'], label='Training', color='#FF533D')
  axes[1].plot(range(stop), hist['val_loss'], label='Validation', color='#03507E')
  axes[1].set title('Loss')
  axes[1].set ylabel('Loss')
  axes[1].set_xlabel('Epoch')
  axes[1].legend(loc='upper right')
  plt.tight_layout();
  print("Best Model:")
  print_best_model_results(hist)
                                                                                                            In [4]:
def plot_acc_metrics(hist1, hist2, stop=50):
  fig, axes = plt.subplots(nrows=2, ncols=1, figsize=(4.25,6))
```

```
axes = axes.flatten()
  axes[0].plot(range(stop), hist1['acc'], label='Training', color='#FF533D')
  axes[0].plot(range(stop), hist1['val_acc'], label='Validation', color='#03507E')
  axes[0].set title('Training')
  axes[0].set ylabel('Accuracy')
  axes[0].set_xlabel('Epoch')
  axes[0].legend(loc='lower right')
  axes[1].plot(range(stop), hist2['acc'], label='Training', color='#FF533D')
  axes[1].plot(range(stop), hist2['val acc'], label='Validation', color='#03507E')
  axes[1].set title('Fine-tuning')
  axes[1].set_ylabel('Accuracy')
  axes[1].set xlabel('Epoch')
  axes[1].legend(loc='lower right')
  plt.tight_layout();
                                                                                                       In [5]:
def print_best_model_results(model_hist):
  best_epoch = np.argmax(model_hist['val_acc'])
  print('epoch:', best_epoch+1, \
  ', val acc:', model hist['val acc'][best epoch], \
  ', val_loss:', model_hist['val_loss'][best_epoch])
                                                                                                       In [6]:
def save bottleneck features():
  datagen = ImageDataGenerator(rescale=1./255)
  model = VGG16(include top=False, weights='imagenet')
  generator = datagen.flow_from_directory(train_data_dir, target_size=(img_width, img_height),
batch_size=batch_size, class_mode=None, shuffle=False)
  bottleneck features train = model.predict generator(generator, nb train samples // batch size)
  np.save(location+'/bottleneck features train.npy',
bottleneck_features_train)
  generator = datagen.flow_from_directory(validation_data_dir, target_size=(img_width, img_height),
batch_size=batch_size, class_mode=None, shuffle=False)
  bottleneck_features_validation = model.predict_generator(generator, nb_validation_samples //
batch size)
  np.save(location+'/bottleneck_features_validation.npy', bottleneck_features_validation)
                                                                                                       In [7]:
def train top model():
  train data = np.load(location+'/bottleneck features train.npy')
```

```
train labels = np.array([0] * (nb train samples // 2) + [1] *
(nb_train_samples // 2))
  validation_data =
np.load(location+'/bottleneck_features_validation.npy')
  validation_labels = np.array([0] * (nb_validation_samples // 2) + [1]
* (nb_validation_samples // 2))
  model = Sequential()
  model.add(Flatten(input shape=train data.shape[1:]))
  model.add(Dense(256,activation='relu'))
  model.add(Dropout(0.5))
  model.add(Dense(1,activation='sigmoid'))
  model.compile(optimizer='rmsprop', loss='binary_crossentropy', metrics=['accuracy'])
  checkpoint = ModelCheckpoint(top model weights path, monitor='val acc', verbose=1,
save_best_only=True, save_weights_only=True, mode='auto')
  fit = model.fit(train_data, train_labels, epochs=epochs,
batch_size=batch_size,validation_data=(validation_data,validation_labels), callbacks=[checkpoint])
  with open(location+'/top_history.txt', 'w') as f:
    json.dump(fit.history, f)
  return model, fit.history
                                                                                                    In [8]:
def finetune_binary_model():
  base_model = VGG16(weights='imagenet', include_top=False,
input shape=(256,256,3)) print("Model loaded.")
  top_model = Sequential()
  top model.add(Flatten(input shape=base model.output shape[1:]))
  top model.add(Dense(256, activation='relu'))
  top model.add(Dropout(0.5))
  top_model.add(Dense(1, activation='sigmoid'))
  top_model.load_weights(top_model_weights_path)
  model = Model(inputs=base model.input,
outputs=top_model(base_model.output))
  for layer in model.layers[:25]:
    layer.trainable = False
  model.compile(loss='binary crossentropy',
optimizer=optimizers.SGD(Ir=1e-4, momentum=0.9), metrics=['accuracy'])
```

```
train datagen = ImageDataGenerator(rescale = 1./255, zoom range=0.2, shear range=0.2,
horizontal_flip=True)
  test datagen = ImageDataGenerator(rescale=1./255)
  train_generator = train_datagen.flow_from_directory(train_data_dir, target_size=(img_height, img_width),
batch_size=batch_size, class_mode='binary')
  validation generator =
test datagen.flow from directory(validation data dir, target size=(img height,
img_width), batch_size=batch_size, class_mode='binary')
  checkpoint = ModelCheckpoint(fine_tuned_model_path, monitor='val_acc', verbose=1,
save_best_only=True, save_weights_only=False, mode='auto')
  fit = model.fit generator(train generator,
steps per epoch=nb train samples//batch size, epochs=epochs,
validation_data=validation_generator,
validation_steps=nb_validation_samples//batch_size, verbose=1, callbacks=[checkpoint])
  with open(location+'/ft_history.txt', 'w') as f:
    ison.dump(fit.history, f)
  return model, fit.history
                                                                                                      In [9]:
def evaluate binary model(model, directory, labels):
  datagen = ImageDataGenerator(rescale=1./255)
  generator = datagen.flow_from_directory(directory,
target_size=(img_height,img_width), batch_size=batch_size, class_mode='binary',
shuffle=False)
  predictions = model.predict generator(generator, len(labels))
  pred labels = [0 if i<0.5 else 1 for i in predictions]
  print(")
  print(classification_report(validation_labels, pred_labels))
  cm = confusion matrix(validation labels, pred labels) return cm
Defining input data
                                                                                                    In [10]:
location = 'data2' top_model_weights_path = location+'/top_model_weights.h5' fine_tuned_model_path =
location+'/ft_model.h5' train_data_dir = location+'/training'
validation data dir = location+'/validation' train samples = [len(os.listdir(train data dir+'/'+i)) for i in
sorted(os.listdir(train data dir))] nb train samples = 1824 validation samples =
[len(os.listdir(validation_data_dir+'/'+i)) for i in sorted(os.listdir(validation_data_dir))]
nb_validation_samples = 448
```

```
img width, img height = 256,256 epochs =
50 batch_size = 16
                                                                                       In []:
save bottleneck features()
                                                                                     In [11]:
d2_model1, d2_history1 = train_top_model()
WARNING:tensorflow:From C:\Anaconda3\envs\envdlcv\lib\site-
packages\tensorflow\python\framework\op_def_library.py:263: colocate_with (from
tensorflow.python.framework.ops) is deprecated and will be removed in a future version. Instructions for
updating:
Colocations handled automatically by placer.
WARNING:tensorflow:From C:\Anaconda3\envs\envdlcv\lib\site-
packages\keras\backend\tensorflow backend.py:3445: calling dropout (from
tensorflow.python.ops.nn_ops) with keep_prob is deprecated and will be removed in a future version.
Instructions for updating:
Please use 'rate' instead of 'keep prob'. Rate should be set to 'rate = 1 - keep prob'.
WARNING:tensorflow:From
C:\Anaconda3\envs\envdlcv\lib\sitepackages\tensorflow\python\ops\math_ops.py:3066: to_int32
(from tensorflow.python.ops.math_ops) is deprecated and will be removed in a future version.
Instructions for updating:
Use tf.cast instead.
Train on 1824 samples, validate on 448 samples
Epoch 1/50
1824/1824 [===============] - 19s 10ms/step - loss: 7.9614
- acc: 0.5016 - val_loss: 8.0590 - val_acc: 0.5000
Epoch 00001: val acc improved from -inf to 0.50000, saving model to data2/top model weights.h5
Epoch 2/50
- acc: 0.5000 - val loss: 8.0590 - val acc: 0.5000
Epoch 00002: val_acc did not improve from 0.50000
Epoch 3/50
- acc: 0.5000 - val_loss: 8.0590 - val_acc: 0.5000
Epoch 00003: val_acc did not improve from 0.50000
Epoch 4/50
- acc: 0.5000 - val_loss: 8.0590 - val_acc: 0.5000
```

Epoch 00004: val acc did not improve from 0.50000

```
Epoch 5/50
- acc: 0.6552 - val_loss: 0.5081 - val_acc: 0.8036
Epoch 00005: val acc improved from 0.50000 to 0.80357, saving model to data2/top model weights.h5
Epoch 6/50
1824/1824 [======================] - 19s 11ms/step - loss: 0.7258
- acc: 0.8026 - val_loss: 0.4214 - val_acc: 0.8549
Epoch 00006: val_acc improved from 0.80357 to 0.85491, saving model to data2/top_model_weights.h5
Epoch 7/50
- acc: 0.8520 - val loss: 0.2513 - val acc: 0.9174
Epoch 00007: val_acc improved from 0.85491 to 0.91741, saving model to data2/top_model_weights.h5
Epoch 8/50
- acc: 0.8739 - val_loss: 0.5095 - val_acc: 0.8460
Epoch 00008: val acc did not improve from 0.91741
Epoch 9/50
- acc: 0.8964 - val_loss: 0.2074 - val_acc: 0.9375
Epoch 00009: val_acc improved from 0.91741 to 0.93750, saving model to data2/top_model_weights.h5
Epoch 10/50
1824/1824 [======================] - 21s 12ms/step - loss: 0.2781
- acc: 0.9052 - val_loss: 0.2311 - val_acc: 0.9286
Epoch 00010: val_acc did not improve from 0.93750
Epoch 11/50
1824/1824 [=============================] - 21s 12ms/step - loss: 0.2184
- acc: 0.9216 - val_loss: 0.2545 - val_acc: 0.9286
Epoch 00011: val acc did not improve from 0.93750
Epoch 12/50
- acc: 0.9189 - val_loss: 0.4140 - val_acc: 0.8728
Epoch 00012: val_acc did not improve from 0.93750
Epoch 13/50
```

```
- acc: 0.9430 - val loss: 0.3403 - val acc: 0.9107
Epoch 00013: val acc did not improve from 0.93750
Epoch 14/50
1824/1824 [======================] - 19s 11ms/step - loss: 0.1734
- acc: 0.9419 - val_loss: 0.2575 - val_acc: 0.9286
Epoch 00014: val acc did not improve from 0.93750
Epoch 15/50
- acc: 0.9501 - val_loss: 0.2354 - val_acc: 0.9330
Epoch 00015: val acc did not improve from 0.93750
Epoch 16/50
1824/1824 [===============] - 19s 10ms/step - loss: 0.0997
- acc: 0.9649 - val_loss: 0.7065 - val_acc: 0.8616
Epoch 00016: val_acc did not improve from 0.93750
Epoch 17/50
- acc: 0.9644 - val loss: 0.3953 - val acc: 0.9263
Epoch 00017: val acc did not improve from 0.93750
Epoch 18/50
1824/1824 [===============] - 18s 10ms/step - loss: 0.1124
- acc: 0.9660 - val_loss: 0.3622 - val_acc: 0.9286
Epoch 00018: val_acc did not improve from 0.93750
Epoch 19/50
1824/1824 [==============] - 18s 10ms/step - loss: 0.0781
- acc: 0.9770 - val_loss: 0.3651 - val_acc: 0.9263
Epoch 00019: val acc did not improve from 0.93750
Epoch 20/50
- acc: 0.9731 - val_loss: 0.7346 - val_acc: 0.8795
Epoch 00020: val_acc did not improve from 0.93750
Epoch 21/50
1824/1824 [===============] - 19s 10ms/step - loss: 0.0980
- acc: 0.9742 - val_loss: 0.6882 - val_acc: 0.8906
```

Epoch 00021: val acc did not improve from 0.93750

```
Epoch 22/50
- acc: 0.9775 - val_loss: 0.4760 - val_acc: 0.9196
Epoch 00022: val acc did not improve from 0.93750
Epoch 23/50
- acc: 0.9825 - val_loss: 0.4074 - val_acc: 0.9375
Epoch 00023: val_acc did not improve from 0.93750
Epoch 24/50
- acc: 0.9819 - val loss: 0.6060 - val acc: 0.9241
Epoch 00024: val_acc did not improve from 0.93750
Epoch 25/50
- acc: 0.9825 - val_loss: 0.4872 - val_acc: 0.9330
Epoch 00025: val acc did not improve from 0.93750
Epoch 26/50
- acc: 0.9836 - val_loss: 0.6003 - val_acc: 0.9152
Epoch 00026: val_acc did not improve from 0.93750
Epoch 27/50
1824/1824 [=====================] - 19s 10ms/step - loss: 0.0808 - acc: 0.9836 - val_loss:
0.3693 - val acc: 0.9241
Epoch 00027: val acc did not improve from 0.93750
Epoch 28/50
- acc: 0.9907 - val_loss: 0.6494 - val_acc: 0.9040
Epoch 00028: val acc did not improve from 0.93750
Epoch 29/50
1824/1824 [=============] - 20s 11ms/step - loss: 0.0501
- acc: 0.9857 - val_loss: 0.4839 - val_acc: 0.9308
Epoch 00029: val acc did not improve from 0.93750
Epoch 30/50
```

```
- acc: 0.9846 - val loss: 0.6352 - val acc: 0.9263
Epoch 00030: val acc did not improve from 0.93750
Epoch 31/50
- acc: 0.9868 - val_loss: 0.4939 - val_acc: 0.9330
Epoch 00031: val acc did not improve from 0.93750
Epoch 32/50
1824/1824 [===============] - 20s 11ms/step - loss: 0.0333
- acc: 0.9901 - val_loss: 0.5689 - val_acc: 0.9286
Epoch 00032: val acc did not improve from 0.93750
Epoch 33/50
- acc: 0.9890 - val_loss: 0.6067 - val_acc: 0.9308
Epoch 00033: val_acc did not improve from 0.93750
Epoch 34/50
- acc: 0.9901 - val loss: 0.5569 - val acc: 0.9241
Epoch 00034: val acc did not improve from 0.93750
Epoch 35/50
- acc: 0.9940 - val_loss: 0.5417 - val_acc: 0.9241
Epoch 00035: val acc did not improve from 0.93750
Epoch 36/50
- acc: 0.9907 - val loss: 0.5860 - val acc: 0.9085
Epoch 00036: val_acc did not improve from 0.93750
Epoch 37/50
- acc: 0.9901 - val loss: 0.5715 - val acc: 0.9286
Epoch 00037: val_acc did not improve from 0.93750
Epoch 38/50
- acc: 0.9890 - val_loss: 0.6733 - val_acc: 0.9129
```

```
Epoch 00038: val acc did not improve from 0.93750
Epoch 39/50
- acc: 0.9890 - val_loss: 0.5863 - val_acc: 0.9219
Epoch 00039: val_acc did not improve from 0.93750
Epoch 40/50
- acc: 0.9951 - val loss: 0.7194 - val acc: 0.9219
Epoch 00040: val acc did not improve from 0.93750
Epoch 41/50
1824/1824 [==============] - 19s 10ms/step - loss: 0.0151
- acc: 0.9956 - val_loss: 0.6697 - val_acc: 0.9174
Epoch 00041: val_acc did not improve from 0.93750
Epoch 42/50
1824/1824 [==============] - 18s 10ms/step - loss: 0.0786
- acc: 0.9857 - val_loss: 0.5607 - val_acc: 0.9174
Epoch 00042: val acc did not improve from 0.93750
Epoch 43/50
- acc: 0.9973 - val loss: 0.6423 - val acc: 0.9286
Epoch 00043: val_acc did not improve from 0.93750
Epoch 44/50
- acc: 0.9901 - val_loss: 0.6561 - val_acc: 0.9174
Epoch 00044: val acc did not improve from 0.93750
Epoch 45/50
1824/1824 [=============================] - 21s 11ms/step - loss: 0.0154
- acc: 0.9940 - val_loss: 0.6484 - val_acc: 0.9241
Epoch 00045: val_acc did not improve from 0.93750
Epoch 46/50
1824/1824 [=================] - 21s 11ms/step - loss: 0.0235
- acc: 0.9956 - val_loss: 0.6600 - val_acc: 0.9241
Epoch 00046: val_acc did not improve from 0.93750
Epoch 47/50
```

- acc: 0.9934 - val loss: 0.7059 - val acc: 0.9219

Epoch 00047: val_acc did not improve from 0.93750

Epoch 48/50

1824/1824 [============] - 18s 10ms/step - loss: 0.0259

- acc: 0.9951 - val_loss: 0.9661 - val_acc: 0.8973

Epoch 00048: val_acc did not improve from 0.93750

Epoch 49/50

1824/1824 [==============] - 18s 10ms/step - loss: 0.0169

- acc: 0.9956 - val_loss: 0.6273 - val_acc: 0.9219

Epoch 00049: val_acc did not improve from 0.93750

Epoch 50/50

1824/1824 [===============] - 18s 10ms/step - loss: 0.0154 - acc: 0.9967 - val_loss:

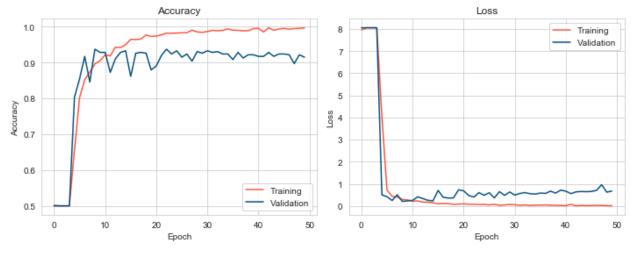
0.6788 - val_acc: 0.9152

Epoch 00050: val_acc did not improve from 0.93750

In [12]:

plot_metrics(d2_history1) Best Model:

epoch: 9, val_acc: 0.9375, val_loss: 0.2073782096683447



Fine Tuning

In []:

ft model, ft history = finetune binary model() Model loaded.

Found 1824 images belonging to 2 classes.

Found 448 images belonging to 2 classes.

Epoch 1/50

114/114 [==============] - 1571s 14s/step - loss: 0.3264 - acc: 0.8799 - val_loss: 0.2074 - val_acc: 0.9375

```
Epoch 00001: val acc improved from -inf to 0.93750, saving model to data2/ft model.h5
Epoch 2/50
val acc: 0.9375
Epoch 00002: val_acc did not improve from 0.93750
Epoch 3/50
- acc: 0.8860 - val_loss: 0.2074 - val_acc: 0.9375
Epoch 00003: val acc did not improve from 0.93750
Epoch 4/50
- acc: 0.8854 - val_loss: 0.2074 - val_acc: 0.9375
Epoch 00004: val_acc did not improve from 0.93750
Epoch 5/50
val acc: 0.9375
Epoch 00005: val acc did not improve from 0.93750
Epoch 6/50
val acc: 0.9375
Epoch 00006: val acc did not improve from 0.93750
Epoch 7/50
val_acc: 0.9375
Epoch 00007: val acc did not improve from 0.93750
Epoch 8/50
val acc: 0.9375
Epoch 00008: val acc did not improve from 0.93750
Epoch 9/50
In [ ]: plot_metrics(ft_history)
Load Model
                                                In [17]:
```

ft_model = load_model(location+'/ft_model.h5')

```
WARNING:tensorflow:From
```

Instructions for updating:

Use tf.cast instead.

C:\Anaconda3\envs\envdlcv\lib\site-packages\keras\engine\saving.py:327: UserWarning: Error in loading the saved optimizer state. As a result, your model is starting with a freshly initialized optimizer.

warnings.warn('Error in loading the saved optimizer'

```
In []:
with open('data1a/top history.txt') as f:
  top_history = json.load(f)
                                                                                                       In []:
with open('data1a/ft_history.txt') as f:
  ft_history = json.load(f)
                                                                                                       In [ ]:
plot_acc_metrics(top_history, ft_history)
                                                                                                    In [22]:
validation_labels = np.array([0] * (nb_validation_samples // 2) + [1] * (nb_validation_samples // 2))
                                                                                                    In [51]:
cm = evaluate_binary_model(ft_model, validation_data_dir, validation_labels)
Found 448 images belonging to 2 classes.
KeyboardInterrupt
                                Traceback (most recent call last)
<ipython-input-51-bf52512d511d> in <module>
----> 1 cm = evaluate_binary_model(ft_model, validation_data_dir, validation_labels)
<ipython-input-27-304db6f68ef2> in evaluate_binary_model(model, directory, labels)
   4 generator = datagen.flow from directory(directory,
target_size=(img_height,img_width), batch_size=batch_size, class_mode='binary',
shuffle=False)
----> 6 predictions = model.predict generator(generator, len(labels))
      pred labels = [0 if i<0.5 else 1 for i in predictions]
C:\Anaconda3\envs\envdlcv\lib\site-packages\keras\legacy\interfaces.py in wrapper(*args, **kwargs)
        warnings.warn('Update your `' + object name + '` call to the ' +
90
        'Keras 2 API: ' + signature, stacklevel=2)
---> 91
             return func(*args, **kwargs)
92
                wrapper. original function = func
93
                return wrapper
```

```
C:\Anaconda3\envs\envdlcv\lib\site-packages\keras\engine\training.py in predict generator(self,
generator, steps, max_queue_size, workers, use_multiprocessing, verbose)
1520
                        workers=workers,
1521
                        use multiprocessing=use multiprocessing,
-> 1522
              verbose=verbose)
C:\Anaconda3\envs\envdlcv\lib\sitepackages\keras\engine\training_generator.py in
predict_generator(model, generator, steps, max_queue_size, workers, use_multiprocessing, verbose)
  451
              x = generator output
  452
--> 453
             outs = model.predict_on_batch(x)
  454
            outs = to_list(outs)
  455
C:\Anaconda3\envs\envdlcv\lib\site-packages\keras\engine\training.py in predict_on_batch(self, x)
1272
1273
                        self. make predict function()
-> 1274
           outputs = self.predict function(ins)
 1275
           return unpack_singleton(outputs)
 1276
C:\Anaconda3\envs\envdlcv\lib\site-
packages\keras\backend\tensorflow_backend.py in call (self, inputs)
 2713
               return self._legacy_call(inputs)
 2714
-> 2715
              return self._call(inputs)
2716
2717
                        if py any(is tensor(x) for x in inputs):
C:\Anaconda3\envs\envdlcv\lib\site-
packages\keras\backend\tensorflow_backend.py in call(self, inputs) 2673
                                                                                 fetched =
self. callable fn(*array vals, run metadata=self.run metadata)
 2674
           else:
-> 2675
              fetched = self._callable_fn(*array_vals)
 2676
           return fetched[:len(self.outputs)]
 2677
C:\Anaconda3\envs\envdlcv\lib\site-
packages\tensorflow\python\client\session.py in __call__(self, *args,
**kwargs)
1437
                     ret = tf_session.TF_SessionRunCallable(
1438
                     self._session._session, self._handle, args, status, -> 1439
                                                                                   run_metadata_ptr)
                     1440
                              if run_metadata:
```

```
1441
            proto data = tf session.TF GetBuffer(run metadata ptr) KeyboardInterrupt:
                                                                                                     In []:
heatmap laebls = ['Damaged', 'Whole']
                                                                                                     In [ ]:
sns.heatmap(cm, annot=True, annot_kws={"size":16}, fmt='g', cmap='OrRd', xticklabels=heatmap_labels,
yticklabels=heatmap labels)
                                                                                                     In [ ]:
sns.heatmap(cm, annot=Ture, annot_kws={"size":16}, fmt='g', cmap='Blues', xticklabels=heatmap_labels,
yticklabels=heatmap labels)
Pipe2
                                                                                                   In [11]:
def pipe2(image_path, model):
  urllib.request.urlretrieve(image_path, 'save.jpg')
  img = load_img('save.jpg', target_size=(256,256))
  x = img_to_array(img)
  x = x.reshape((1,) + x.shape)/255
  pred = model.predict(x)
  print("Validating that damage exists....")
  print(pred)
  if(pred[0][0]<=0.5):
    print("Validation complete - proceed to location and severity determination")
    print ("Are you sure that your car is damaged? Please submit another picture of the damage.")
    print ("Hint: Try zooming in/out, using a different angle or different lighting")
                                                                                                   In [12]:
Image('http://3.bp.blogspot.com/-
PrRY9XxCqYQ/UDNutnMI7LI/AAAAAAAAAddw/UGygghh-hRA/s1600/Bumper+scuff.JPG')
```

Out[12]:



In [18]: pipe2('http://3.bp.blogspot.com/-

 $\label{lem:prryyxxcqyQ/UDNutnMI7LI/AAAAAAAAAAAABdw/UGygghh-hRA/s1600/Bumper+scuff.JPG', ft_model) \\ Validating that damage exists....$

[[0.0002488]]

Validation complete - proceed to location and severity determination

In [40]:

Image('https://i.ytimg.com/vi/4oV1klVPogY/maxresdefault.jpg')

Out[40]:



In [41]:

pipe2('https://i.ytimg.com/vi/4oV1klVPogY/maxresdefault.jpg', ft_model) Validating that damage exists.... [[0.01300194]]

Validation complete - proceed to location and severity determination

In [47]:

Image('http://blog.automart.co.za/wpcontent/uploads/2014/09/Accident_Damaged_Car.png')

Out[47]:



In [46]: ft_model)

pipe2('http://blog.automart.co.za/wpcontent/uploads/2014/09/Accident_Damaged_Car.png', Validating that damage exists....

[[0.11757535]]

Validation complete - proceed to location and severity determination