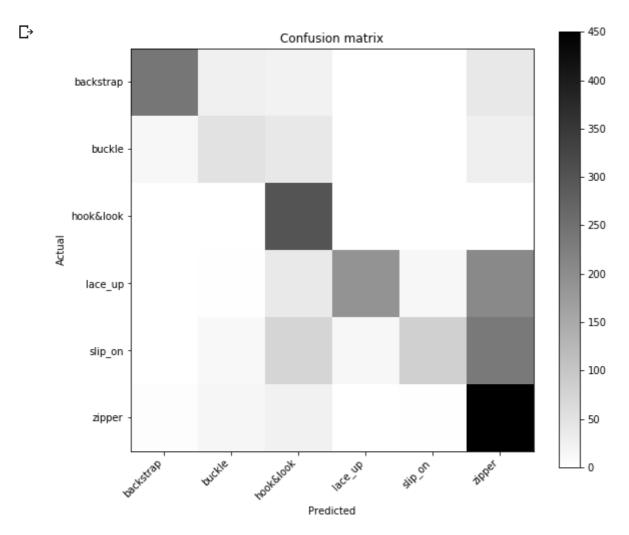
```
from keras.preprocessing.image import img_to_array
from keras models import load model
import numpy as np
import pandas as pd
import argparse
import imutils
import pickle
import cv2
import os
from keras.applications.vgg19 import VGG19
from keras.models import Model
from keras.layers import Dense, GlobalAveragePooling2D, Dropout
from keras.optimizers import SGD
import matplotlib.pyplot as plt
import urllib.request
   Using TensorFlow backend.
data list = pd.read excel('Internship data.xlsx')
# Run this cell to mount your Google Drive.
from google colab import drive
drive.mount('/content/drive')
!unzip drive/My\ Drive/final data-fynd.zip
labels = ['backstrap BV',
 'backstrap NBV',
 'buckle BV
 'buckle_NBV'
 'hook&look_BV'
 'hook&look_NBV'
 'lace up B\overline{V}'
 'lace_up_NBV'
 'slip on BV'
 'slip_on_NBV',
 'zipper_BV'
 'zipper_NBV']
vgg_model = VGG19(weights='imagenet', include_top=False)
   WARNING: Logging before flag parsing goes to stderr.
    W0630 10:36:18.381148 139780937172864 deprecation_wrapper.py:119] From /usr/l
    W0630 10:36:18.402068 139780937172864 deprecation_wrapper.py:119] From /usr/l
    W0630 10:36:18.405764 139780937172864 deprecation_wrapper.py:119] From /usr/l
    W0630 10:36:18.436507 139780937172864 deprecation wrapper.py:119] From /usr/l
    W0630 10:36:18.984603 139780937172864 deprecation wrapper.py:119] From /usr/l
    W0630 10:36:18.986264 139780937172864 deprecation wrapper.py:119] From /usr/l
```

```
x = vqq model.output
x = GlobalAveragePooling2D()(x)
# add fully-connected layer
x = Dense(512, activation='relu')(x)
x = Dropout(0.3)(x)
# add output layer
predictions = Dense(12, activation='softmax')(x)
model = Model(inputs=vgg model.input, outputs=predictions)
model.load weights('drive/My Drive/fine tune shoes multiclass.best.hdf5')
    W0630 10:36:21.114876 139780937172864 deprecation.py:506] From /usr/local/lib
     Instructions for updating:
     Please use `rate` instead of `keep prob`. Rate should be set to `rate = 1 - k
def identify type submission(images , labels):
    class_link = pd.DataFrame(columns = ['predicted_class', 'best view image'])
    for i in range(0,len(images['class'])):
      best_image_index = 0
      type score = dict()
      image list = []
      best \overline{i}mage link = []
      j = \overline{0}
      for view in ['view 1', 'view 2', 'view 3', 'view 4', 'view 5']:
        try:
          url_response = urllib.request.urlopen(images[view][i])
          img_array = np.array(bytearray(url_response.read()), dtype=np.uint8)
          image = cv2.imdecode(img_array, -1)
          image_list.append(image)
          best_image_link.append(images[view][i])
          image = cv2.resize(image, (128,128))
          image = img to array(image)
          image = np.expand_dims(image, axis=0)
          proba = model.predict(image)[0]
          idxs = np.argsort(proba)[::-1][:1]
          type_score[j] = {proba[idxs[0]] : labels[idxs[0]]}
j = j+1
        except:
          continue
      best image = dict()
      count = 0
      for key,i in type_score.items() :
    type_image = list(i.values())[0]
          score = list(i.keys())[0]
if type_image.split('_')[1] == 'BV':
               best image[count] = {score : type_image.split('_')[0]}
          count = count + 1
      type_of_footwear = ''
      if len(best_image) == 1:
        for key,value in best_image.items():
               type_of_footwear = list(value.values())[0]
```

```
best image index = list(best image.keys())[0]
        class link = class_link.append(pd.Series([type_of_footwear, best_image_link[best
      if len(best_image) > 1 :
        \max \ \text{score} = 0
        best image index = 0
        for key,value in best image.items():
          if list(value.keys())[0] > max_score:
            max score = list(value.keys())[0]
             best_image_index = key
             type of footwear = list(value.values())[0]
        class \lim_{n \to \infty} \frac{1}{n} = \frac{1}{n} \ln n. Series([type of footwear, best image \lim_{n \to \infty} \frac{1}{n} \ln n]
      if len(best image) == 0:
        \max score = 0
        bes\overline{t} image index = 0
        for key,value in type_score.items():
   if list(value.keys())[0] > max_score:
            max_score = list(value.keys())[0]
            best_image_index = key
            type_of_footwear = list(value.values())[0].split(' ')[0]
        class link = class link.append(pd.Series([type of footwear, best image link[best
    return class link
%%time
predicted data list = identify type submission(data list, labels)
CPU times: user 1h 15min 16s, sys: 57.8 s, total: 1h 16min 14s
    Wall time: 6h 53min 45s
predicted_data_list.loc[predicted_data_list['predicted_class'] == 'slip', 'predicted_cla
predicted_data_list.loc[predicted_data_list['predicted_class'] == 'lace', 'predicted_cla
new data = pd.concat([data list, predicted data list], axis=1)
new_data.to_excel('Internship_Data-Predicted(Final).xlsx', index=False)
!pip install pandas ml
Collecting pandas ml
       Downloading <a href="https://files.pythonhosted.org/packages/ae/72/6d90debfcb9ea74ec">https://files.pythonhosted.org/packages/ae/72/6d90debfcb9ea74ec</a>
                                                | 102kB 2.7MB/s
     Requirement already satisfied: enum34 in /usr/local/lib/python3.6/dist-packag
     Requirement already satisfied: pandas>=0.19.0 in /usr/local/lib/python3.6/dis
     Requirement already satisfied: pytz>=2011k in /usr/local/lib/python3.6/dist-p
     Requirement already satisfied: python-dateutil>=2.5.0 in /usr/local/lib/pytho
     Requirement already satisfied: numpy>=1.12.0 in /usr/local/lib/python3.6/dist
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.6/dist-pack
     Installing collected packages: pandas-ml
     Successfully installed pandas-ml-0.6.1
from sklearn.metrics import confusion matrix, accuracy score
from pandas ml import ConfusionMatrix
import matplotlib.pyplot as plt
cm = ConfusionMatrix(new data['class'], new data['predicted class'])
cm
C
```

Predicted Actual	backstrap	buckle	hook&look	lace_up	slip_on	zipper	all
					_		
backstrap	241	27	24	0	1	42	335
buckle	15	52	41	1	1	29	139
hook&look	1	0	302	Θ	Θ	1	304
lace_up	0	2	39	189	15	208	453
slip_on	1	13	75	15	84	237	425
zinnar	5	16	26	1	2	15A	500

cm.plot()
plt.show()



accuracy_score(new_data['class'], new_data['predicted_class'])

□→ 0.6113172541743971