

```
In [ ]: from google.colab import drive  
drive.mount("/content/drive")
```

Mounted at /content/drive

```
In [ ]: import os
import cv2
import numpy as np
from PIL import Image
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import tensorflow
from keras.models import Sequential
from keras.layers.convolutional import Conv2D
from keras.layers import BatchNormalization
from keras.layers.convolutional import MaxPooling2D
from keras.preprocessing import image
from keras.layers.core import Activation
from keras.layers.core import Dropout
from keras.layers import Dense
from keras.layers import Flatten
from keras.layers import Input
from keras.models import Model
from keras.applications.imagenet_utils import decode_predictions, preprocess_input
from tensorflow.keras.optimizers import Adam
from keras.models import model_from_json
from sklearn.model_selection import train_test_split
import numpy as np
import pandas as pd
import keras
from keras.models import load_model
import urllib.request
import pickle
import time
from sklearn.decomposition import PCA
import random
from scipy.spatial import distance
import matplotlib.image as mpimg
```

```
In [ ]: os.chdir("/content/drive/MyDrive/eyewear")
```

```
In [ ]: itg_eyewear = pd.read_csv("eyewear_ml_challenge.csv")
        itg_eyewear.head()
```

```
Out[6]:
```

	Unnamed: 0	product_name	product_id	parent_category	Image_Front	frame_shape
0	0	Vintage Persona C4	7641	eyeframe	http://tak-apps-dev.s3.amazonaws.com/recruitme...	Rectangle
1	1	Vintage Crazy-X C2 Red	7643	eyeframe	http://tak-apps-dev.s3.amazonaws.com/recruitme...	Rectangle
2	2	Jialedi Eyewear 1086 C8 Blue	8254	eyeframe	http://tak-apps-dev.s3.amazonaws.com/recruitme...	Rectangle
3	3	Jialedi Eyewear 1086 C6 Red	8255	eyeframe	http://tak-apps-dev.s3.amazonaws.com/recruitme...	Rectangle
4	4	Jialedi Eyewear 1086 C4 Tortoise Transperent	8256	eyeframe	http://tak-apps-dev.s3.amazonaws.com/recruitme...	Rectangle

```
In [ ]: itg_eyewear.describe()
```

```
In [ ]: for i, url in enumerate(itg_eyewear["Image_Front"]):
        try:
            urllib.request.urlretrieve(url, str(i)+".png")
        except:
            pass
```

```
In [ ]: model = tensorflow.keras.applications.VGG16(weights='imagenet', include_top=True)
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5 (https://storage.googleapis.com/tensorflow/keras-applications/vgg16/vgg16_weights_tf_dim_ordering_tf_kernels.h5)

553467904/553467096 [=====] - 13s 0us/step

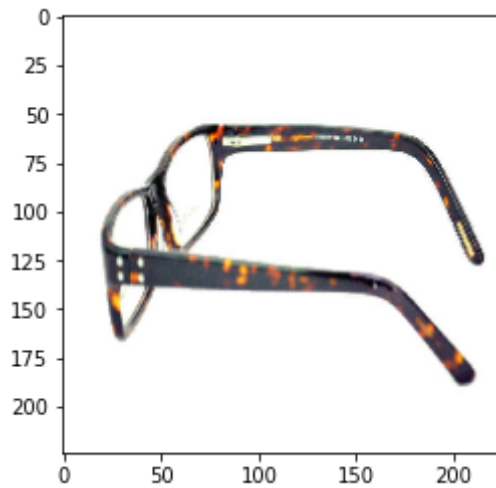
553476096/553467096 [=====] - 13s 0us/step

```
In [ ]: def load_image(path):  
        img = image.load_img(path, target_size=model.input_shape[1:3])  
        x = image.img_to_array(img)  
        x = np.expand_dims(x, axis=0)  
        x = preprocess_input(x)  
        return img, x
```

```
In [ ]: img, x = load_image("0.png")  
        print("shape of x: ", x.shape)  
        print("data type: ", x.dtype)  
        plt.imshow(img)
```

shape of x: (1, 224, 224, 3)
data type: float32

Out[22]: <matplotlib.image.AxesImage at 0x7f28901f0790>



```
In [ ]: predictions = model.predict(x)

# print out the
for _, pred, prob in decode_predictions(predictions)[0]:
    print("predicted %s with probability %0.3f" % (pred, prob))
```

Downloading data from https://storage.googleapis.com/download.tensorflow.org/data/imagenet_class_index.json (https://storage.googleapis.com/download.tensorflow.org/data/imagenet_class_index.json)

40960/35363 [=====] - 0s 0us/step

49152/35363 [=====] - 0s 0us/step

predicted black_widow with probability 0.196

predicted bow with probability 0.120

predicted tricycle with probability 0.089

predicted muzzle with probability 0.084

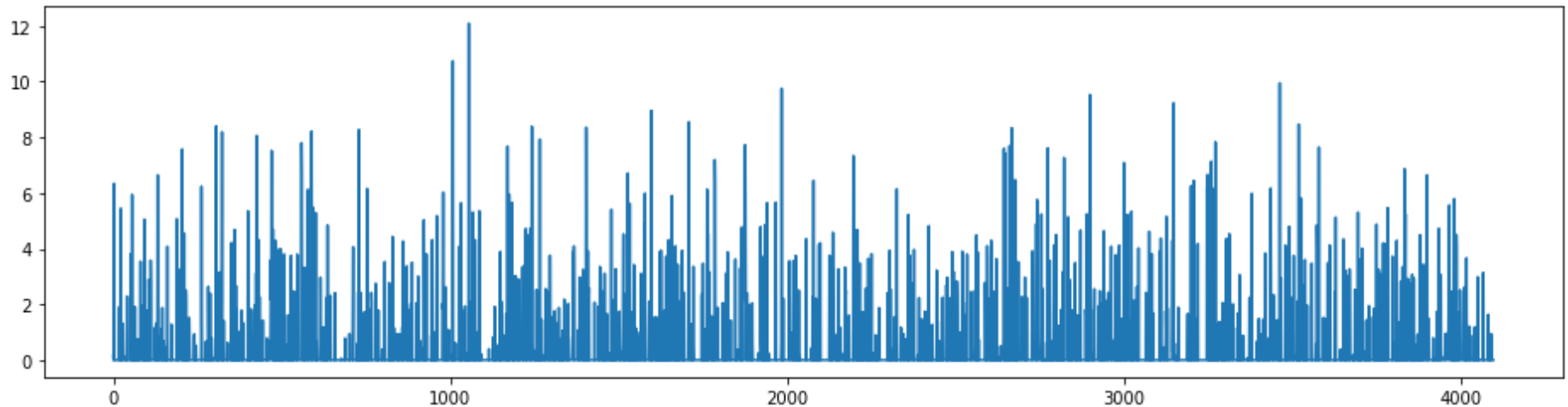
predicted stretcher with probability 0.051

```
In [ ]: feat_extractor = Model(inputs=model.input, outputs=model.get_layer("fc2").output)
        feat_extractor.summary()
```

```
In [ ]: img, x = load_image("0.png")
        feat = feat_extractor.predict(x)

        plt.figure(figsize=(16,4))
        plt.plot(feat[0])
```

Out[25]: [<matplotlib.lines.Line2D at 0x7f281ae821d0>]



```
In [ ]: images_path = '.'
        image_extensions = ['.jpg', '.png', '.jpeg'] # case-insensitive (upper/lower doesn't matter)
        max_num_images = 10000

        images = [os.path.join(dp, f) for dp, dn, filenames in os.walk(images_path) for f in filenames if os.path.splitext(f)[1]
        ...
        if max_num_images < len(images):
            images = [images[i] for i in sorted(random.sample(xrange(len(images)), max_num_images))]
        ...

        print("keeping %d images to analyze" % len(images))
```

keeping 5531 images to analyze

In []:

```
tic = time.clock()

features = []
for i, image_path in enumerate(images):
    if i % 500 == 0:
        toc = time.clock()
        elap = toc-tic;
        print("analyzing image %d / %d. Time: %4.4f seconds." % (i, len(images), elap))
        tic = time.clock()
    try:
        img, x = load_image(image_path);
        feat = feat_extractor.predict(x)[0]
        features.append(feat)
    except:
        pass

print('finished extracting features for %d images' % len(images))
```

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:2: DeprecationWarning: time.clock has been deprecated in Python 3.3 and will be removed from Python 3.8: use time.perf_counter or time.process_time instead

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:8: DeprecationWarning: time.clock has been deprecated in Python 3.3 and will be removed from Python 3.8: use time.perf_counter or time.process_time instead

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:11: DeprecationWarning: time.clock has been deprecated in Python 3.3 and will be removed from Python 3.8: use time.perf_counter or time.process_time instead
This is added back by InteractiveShellApp.init_path()

```
analyzing image 0 / 5531. Time: 0.0004 seconds.
analyzing image 500 / 5531. Time: 52.3520 seconds.
analyzing image 1000 / 5531. Time: 52.7464 seconds.
analyzing image 1500 / 5531. Time: 57.9265 seconds.
analyzing image 2000 / 5531. Time: 54.2869 seconds.
analyzing image 2500 / 5531. Time: 55.6184 seconds.
analyzing image 3000 / 5531. Time: 56.4256 seconds.
analyzing image 3500 / 5531. Time: 54.0333 seconds.
analyzing image 4000 / 5531. Time: 55.2164 seconds.
analyzing image 4500 / 5531. Time: 54.0489 seconds.
```

```
analyzing image 5000 / 5531. Time: 53.5360 seconds.  
analyzing image 5500 / 5531. Time: 53.5516 seconds.  
finished extracting features for 5531 images
```

In []:

```
features = np.array(features)  
pca = PCA(n_components=300)  
pca.fit(features)
```

Out[33]: PCA(copy=True, iterated_power='auto', n_components=300, random_state=None, svd_solver='auto', tol=0.0, whiten=False)

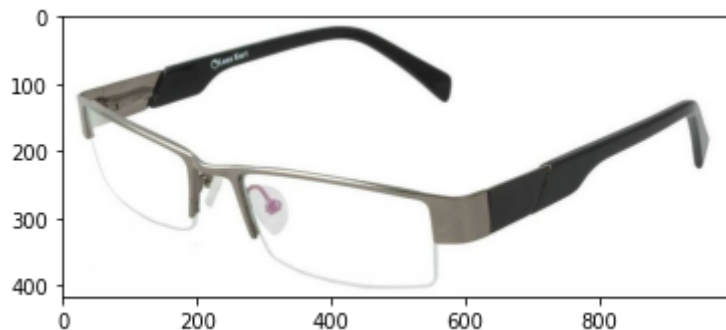
In []:

```
pca_features = pca.transform(features)
```

In []:

```
# grab a random query image  
query_image_idx = int(len(images) * random.random())  
  
# let's display the image  
img = image.load_img(images[query_image_idx])  
plt.imshow(img)
```

Out[12]: <matplotlib.image.AxesImage at 0x7f4220041890>



In []:

```
similar_idx = [ distance.cosine(pca_features[query_image_idx], feat) for feat in pca_features ]
```

In []:

```
idx_closest = sorted(range(len(similar_idx)), key=lambda k: similar_idx[k])[1:6]
```

In []:

```
thumbs = []
for idx in idx_closest:
    img = image.load_img(images[idx])
    img = img.resize((int(img.width * 100 / img.height), 100))
    thumbs.append(img)

# concatenate the images into a single image
concat_image = np.concatenate([np.asarray(t) for t in thumbs], axis=1)

# show the image
plt.figure(figsize = (16,12))
plt.imshow(concat_image)
```

Out[39]: <matplotlib.image.AxesImage at 0x7f27aaf79e50>



In []:

```
def get_closest_images(query_image_idx, num_results=10):
    distances = [ distance.euclidean(pca_features[query_image_idx], feat) for feat in pca_features ]
    idx_closest = sorted(range(len(distances)), key=lambda k: distances[k])[0:num_results]
    return idx_closest

def get_concatenated_images1(indexes, thumb_height):
    thumbs = []
    for idx in indexes:
        img = image.load_img(images[idx])
        img = img.resize((int(img.width * thumb_height / img.height), thumb_height))
        thumbs.append(img)
    concat_image = np.concatenate([np.asarray(t) for t in thumbs], axis=1)
    return concat_image

def get_concatenated_images(indexes, thumb_height):
    thumbs = []
    for idx in indexes[:5]:
        img = image.load_img(images[idx])
        img = img.resize((int(img.width * thumb_height / img.height), thumb_height))
        thumbs.append(img)
    concat_image1 = np.concatenate([np.asarray(t) for t in thumbs], axis=1)
    thumbs=[]
    for idx in indexes[5:10]:
        img = image.load_img(images[idx])
        img = img.resize((int(img.width * thumb_height / img.height), thumb_height))
        thumbs.append(img)
    concat_image2 = np.concatenate([np.asarray(t) for t in thumbs], axis=1)
    return concat_image1 , concat_image2
```

```

In [ ]: query_image_idx = int(len(images) * random.random())
idx_closest = get_closest_images(query_image_idx)
query_image = get_concatenated_images1([query_image_idx], 300)
results_image = get_concatenated_images(idx_closest, 200)

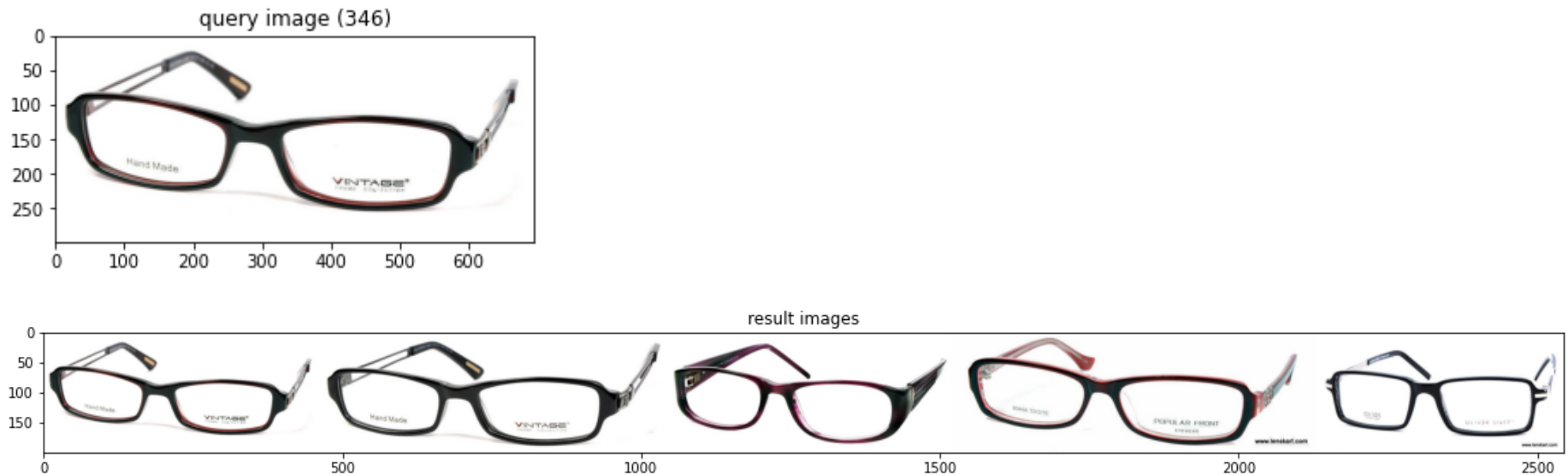
# display the query image
plt.figure(figsize = (5,5))
plt.imshow(query_image)
plt.title("query image (%d)" % query_image_idx)

# display the resulting images
plt.figure(figsize = (20,10))
plt.imshow(results_image[0])
plt.title("result images")

# display the resulting images
plt.figure(figsize = (20,10))
plt.imshow(results_image[1])
plt.title("result images")

```

Out[15]: Text(0.5, 1.0, 'result images')





```
In [ ]: try:
        urllib.request.urlretrieve(input("Enter url: "), "uploaded.png")
    except:
        pass
    # load image and extract features
    new_image, x = load_image("uploaded.png")
    new_features = feat_extractor.predict(x)

    # project it into pca space
    new_pca_features = pca.transform(new_features)[0]

    # calculate its distance to all the other images pca feature vectors
    distances = [ distance.cosine(new_pca_features, feat) for feat in pca_features ]
    idx_closest = sorted(range(len(distances)), key=lambda k: distances[k])[0:10] # grab first 10
    results_image = get_concatenated_images(idx_closest, 200)

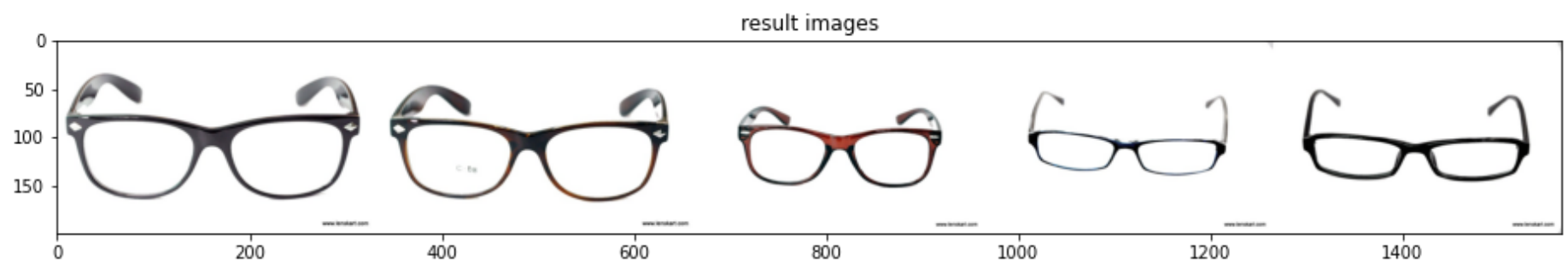
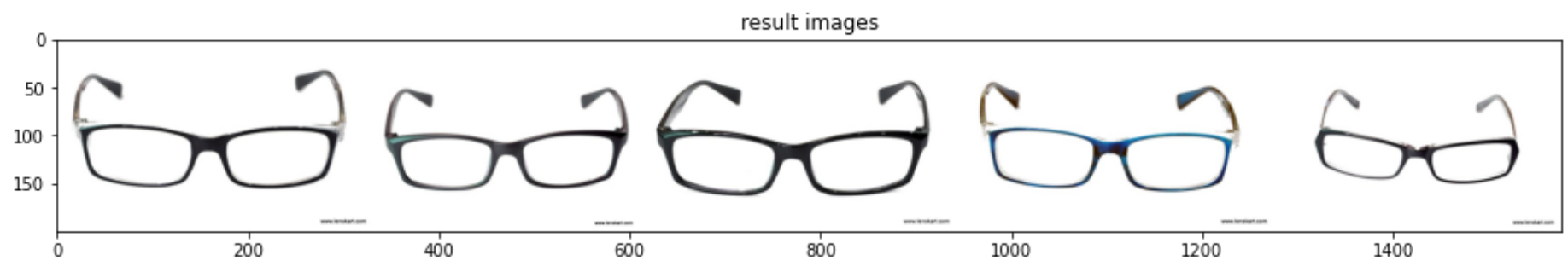
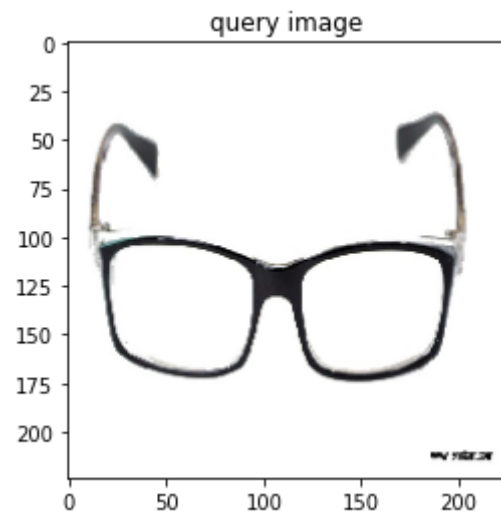
    # display the results
    plt.figure()
    plt.imshow(new_image)
    plt.title("query image")

    # display the resulting images
    plt.figure(figsize = (16,12))
    plt.imshow(results_image[0])
    plt.title("result images")

    # display the resulting images
    plt.figure(figsize = (16,12))
    plt.imshow(results_image[1])
    plt.title("result images")
```

Enter url: http://tak-apps-dev.s3.amazonaws.com/recruitment/machine-learning/dataset/DSC_0049_4.JPG (http://tak-apps-dev.s3.amazonaws.com/recruitment/machine-learning/dataset/DSC_0049_4.JPG)

Out[30]: Text(0.5, 1.0, 'result images')



Task 2

```
In [ ]: query_image_idx = int(len(images) * random.random())
query_image = get_concatenated_images1([query_image_idx], 300)
idx_closest = get_closest_images(query_image_idx, 1000)

plt.figure(figsize = (5,5))
plt.imshow(query_image)
plt.title("query image (%d)" % query_image_idx)

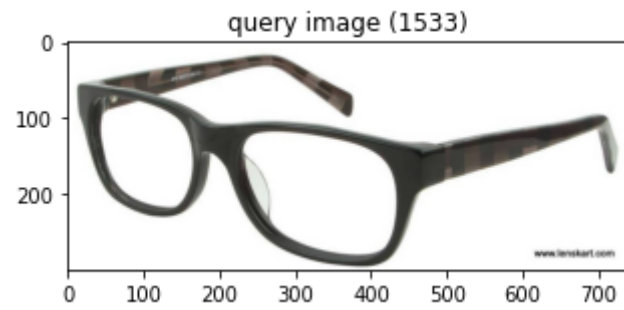
#frame_shape
shapes = []
for id in idx_closest[:7]:
    shapes.append(itg_eyewear['frame_shape'][id])

#parent_category
category = []
for id in idx_closest:
    category.append(itg_eyewear['parent_category'][id])

print("Predicted frame Shape: ", max(shapes, key=shapes.count))
print("Predicted Parent Category: ", max(category, key=category.count))

print("Original frame Shape: ", itg_eyewear['frame_shape'][query_image_idx])
print("Original Parent Category: ", itg_eyewear['parent_category'][query_image_idx])
```

```
Predicted frame Shape: Rectangle
Predicted Parent Category: eyeframe
Original frame Shape: Rectangle
Original Parent Category: eyeframe
```




```

In [ ]: try:
        urllib.request.urlretrieve(input("Enter url: "), "uploaded.png")
    except:
        pass
    # Load image and extract features
    new_image, x = load_image("uploaded.png")
    new_features = feat_extractor.predict(x)

    # project it into pca space
    new_pca_features = pca.transform(new_features)[0]

    # calculate its distance to all the other images pca feature vectors
    distances = [ distance.cosine(new_pca_features, feat) for feat in pca_features ]
    idx_closest = sorted(range(len(distances)), key=lambda k: distances[k])[0:1000] # grab first 1000

    plt.figure()
    plt.imshow(new_image)
    plt.title("query image")

    #frame_shape
    shapes = []
    for id in idx_closest[:7]:
        shapes.append(itg_eyewear['frame_shape'][id])

    #parent_category
    category = []
    for id in idx_closest:
        category.append(itg_eyewear['parent_category'][id])

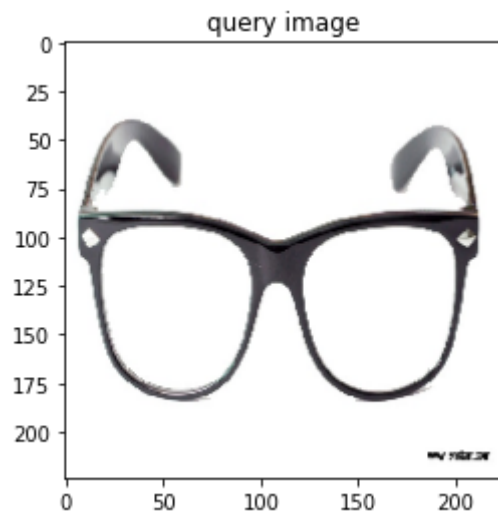
    print("Predicted frame Shape: ", max(shapes, key=lambda k: shapes[k]))
    print("Predicted Parent Category: ", max(category, key=lambda k: category[k]))

```

Enter url: http://tak-apps-dev.s3.amazonaws.com/recruitment/machine-learning/dataset/DSC_0201_3.JPG (http://tak-apps-dev.s3.amazonaws.com/recruitment/machine-learning/dataset/DSC_0201_3.JPG)

Predicted frame Shape: Wayfarer

Predicted Parent Category: eyeframe



```
In [ ]: pickle.dump([images, pca_features, pca], open('model.p', 'wb'))
```

```
In [ ]: file = open("model.p", "rb")  
images, pca_features, pca = pickle.load(file)
```