## **ASSIGNMENT 2**

# **COMPUTER VISION**

## **SUBMITTED BY:**

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**DATED:**13-Dec-2020

**SUBMITTED TO:** 

DR. MOAZZAM FARAZ

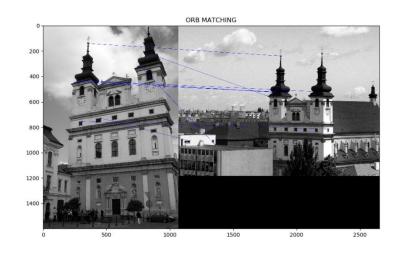
#### TASK1:

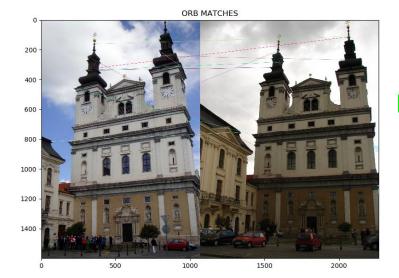
### Feature Mapping in Images Using Feature Detectors and Descriptors

#### 1. ORB:

Top 10 Matches found using ORB are as follows:

Feature matching on building with blue marker





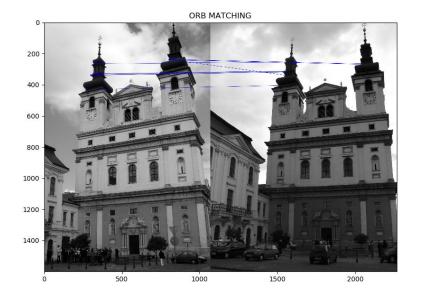
For building 2 and 3 top 10 matches



Top 10 features of book and person holding book

Matches with green marker for roma1 and roma 2



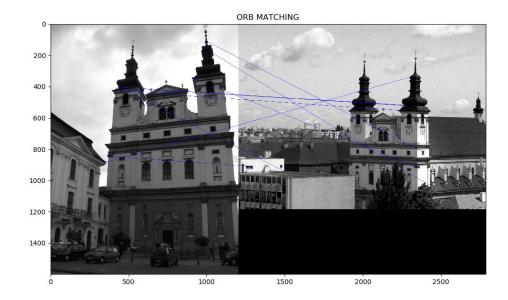


ORB Feature matching in Grascale images



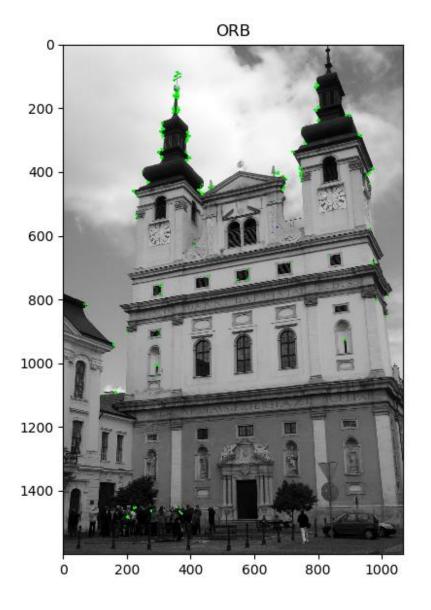
# Blue markers reprensting 10 matches of roma

# For next 2 buildings feature matches



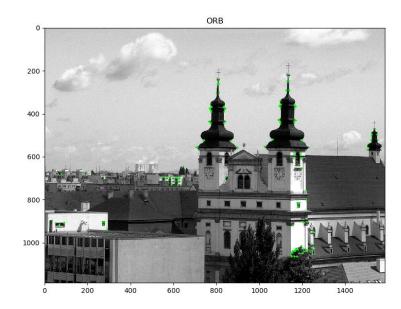


Feature matches for person holding book



Buidling 2 corners detection without matching

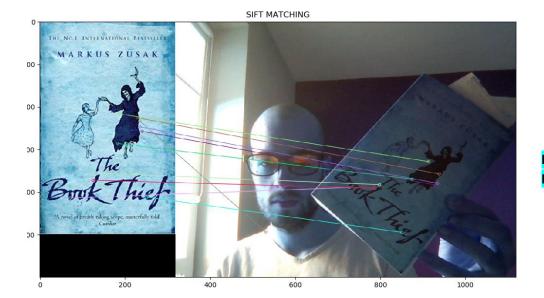
Precise Corner Detected of Buiding 1
Via ORB



```
import cv2
from matplotlib import pyplot as plt
# you can use following code to increase or decrease your figure size
from pylab import rcParams
rcParams['figure.figsize'] = 5, 5
img = cv2.imread('building_2.jpg',0)
img1=cv2.imread('building_3.jpg',0)
print(img.shape)
# Initiate ORB detector
orb = cv2.ORB create()
# find the keypoints and descriptors with ORB
kp orb, des orb = orb.detectAndCompute(img, None)
kp_orb1, des_orb1 = orb.detectAndCompute(img1, None)
len(kp_orb),len(kp_orb1)
# Match descriptors.
# create BFMatcher object
bf = cv2.BFMatcher(cv2.NORM HAMMING, crossCheck=True)
matches = bf.match(des orb, des orb1)
# Sort them in the order of their distance.
matches = sorted(matches, key = lambda x:x.distance)
# Draw first 10 matches.
img3 = cv2.drawMatches(img,kp_orb,img1,kp_orb1,matches[:10],img1,
flags=2,matchColor=(0,0,255))
plt.imshow(img3)
plt.title("ORB MATCHING")
plt.show()
```

#### 2. SIFT

**TOP 10 Features Matches of SIFT** 



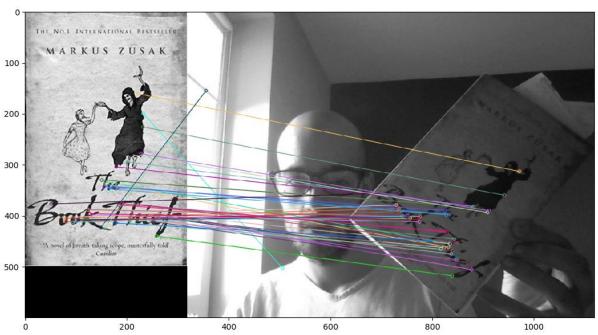
Sift match for Book and Person Holding book

Sift match for RGB book image and person holding book





Feature mapping in two rotated buildings features



```
import cv2
import matplotlib.pyplot as plt
#matplotlib inline
#read images

img2=cv2.imread('book.jpg')
img3=cv2.imread('book_person_holding.jpg')

img2=cv2.cvtColor(img2,cv2.COLOR_BGR2GRAY)
img3=cv2.cvtColor(img3,cv2.COLOR_BGR2GRAY)
#SIFT
sift=cv2.xfeatures2d.SIFT_create()

keypoints_2,descriptors_2=sift.detectAndCompute(img2,None)
keypoints_3,descriptors_3=sift.detectAndCompute(img3,None)
```

```
len (keypoints_2),len (keypoints_3)
bf=cv2.BFMatcher(cv2.NORM_L1,crossCheck=True)
matches=bf.match(descriptors_2,descriptors_3)
matches=sorted(matches,key=lambda x:x.distance)
img4=cv2.drawMatches(img2,keypoints_2,img3,keypoints_3,matches[:10],img3,flags=2)
plt.imshow(img4)
plt.show()
```

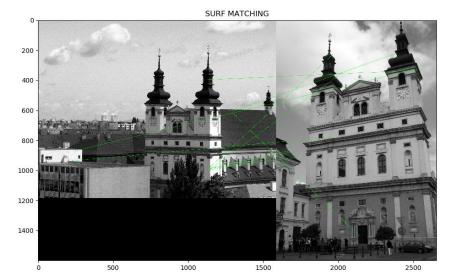
#### 3. SURF



SURF matches for book and person holding book







Grayscale images mapping on two buildings



SURF top 10 features for

2 rotated buildings



**RGB Feature** matches for <mark>roma</mark> **buildings** 



1000

1500

2000

SURF MATCHING

```
SURF matches for RGB
<mark>images</mark>
```

0

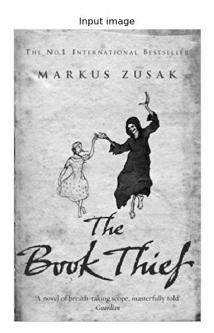
```
import cv2
import matplotlib.pyplot as plt
#matplotlib inline
#read images
img2=cv2.imread('book.jpg')
img3=cv2.imread('book_person_holding.jpg')
img2=cv2.cvtColor(img2,cv2.COLOR BGR2GRAY)
img3=cv2.cvtColor(img3,cv2.COLOR_BGR2GRAY)
#SIFT
surf=cv2.xfeatures2d.SURF_create()
keypoints_2, descriptors_2=surf.detectAndCompute(img2, None)
keypoints 3, descriptors 3=surf.detectAndCompute(img3, None)
```

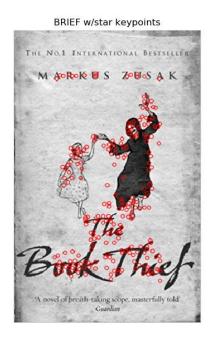
500

```
len (keypoints_2), len (keypoints_3)
bf=cv2.BFMatcher (cv2.NORM_L1, crossCheck=True)
matches=bf.match (descriptors_2, descriptors_3)
matches=sorted (matches, key=lambda x:x.distance)
img4=cv2.drawMatches (img2, keypoints_2, img3, keypoints_3, matches[:10], img3, flags=2)
plt.title("SURF MATCHING")
plt.imshow(img4)
plt.show()
```

#### 4. BRIEF

#### **Feature Detection Using BRIEF Feature Detector**

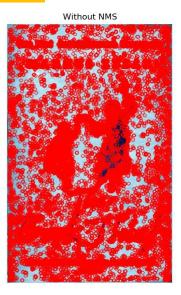




#### 5. FAST

#### FAST Features detected for book





#### FAST Features detected for person holding book





TASK 2:

## **Results of Few Correctly Classified Images**

Positive image 0



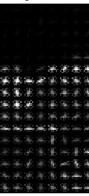
Postive HOG



Negative image 0



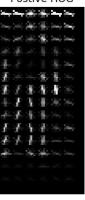
Negative HOG



Positive image 0



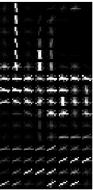
Postive HOG



Negative image 0



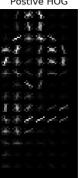
Negative HOG



Positive image 0



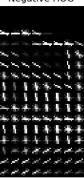
Postive HOG



Negative image 0



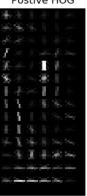
Negative HOG



Positive image 0



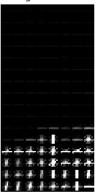
Postive HOG



Negative image 0



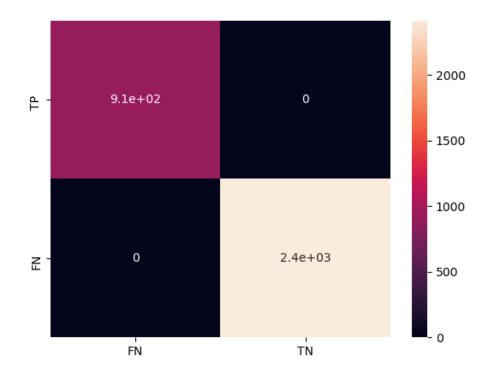
Negative HOG



**Quantitative Performance Measures: (Support Vector Machine Classifier)** 

True Positive Rate: 911 False Positive Rate: 0 True Negative Rate: 2416 False Negative Rate: 0

#### **Confusion Matrix SVM**

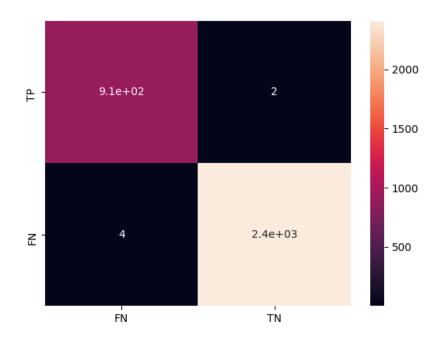


```
Accuracy: 1.0
finish learning SVM.
LinearSVC(C=1.0, class_weight=None, dual=True, fit_intercept=True,
     intercept_scaling=1, loss='squared_hinge', max_iter=1000,
     multi_class='ovr', penalty='l2', random_state=None, tol=0.0001,
    verbose=0)
 trainSVM ×
  warnings.warn(msg, category=DeprecationWarning)
 start learning SVM.
Accuracy: 1.0
finish learning SVM.
LinearSVC(C=1.0, class_weight=None, dual=True, fit_intercept=True,
           intercept_scaling=1, loss='squared_hinge', max_iter=1000,
           multi_class='ovr', penalty='12', random_state=None, to1=0.0001,
           verbose=0)
 1.0
 [[ 911
          0]
 [ 0 2416]]
 Process finished with exit code 0
```

Quantitative Performance Measures: Random Forest Classifier

True Positive Rate: 909
False Positive Rate: 2
True Negative Rate: 2412
False Negative Rate: 4

#### **Confusion Matrix Random Forest Classifier**



# Accuracy: 0.9981965734896303 RandomForestClassifier(bootstrap=True, class\_weight=None, criterion='gini', max\_depth=None, max\_features='auto', max\_leaf\_nodes=None, min\_impurity\_decrease=0.0, min\_impurity\_split=None, min\_samples\_leaf=1, min\_samples\_split=2, min\_weight\_fraction\_leaf=0.0, n\_estimators=10, n\_jobs=None, oob\_score=False, random\_state=0, verbose=0, warm\_start=False)

```
trainRF ×
C:\Users\eshas\AppData\Local\Programs\Python\Python36\python.exe C:/Users/eshas/PycharmProjects/untitled/trainRF.py
start loading 2416 positive files
start loading 911 negative files
(3327, 3780)
(3327,)
Training Random Forest
C:\Users\eshas\AppData\Roaming\Python\Python36\site-packages\sklearn\externals\joblib\ init .py:15: DeprecationWarning:
  warnings.warn(msg, category=DeprecationWarning)
C:\Users\eshas\AppData\Roaming\Python\Python36\site-packages\sklearn\ensemble\forest.py:245: FutureWarning: The default va
  "10 in version 0.20 to 100 in 0.22.", FutureWarning)
Training Complete
Accuracy: 0.9981965734896303
Finish learning Random Forest.
RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                      max_depth=None, max_features='auto', max_leaf_nodes=None,
                      min_impurity_decrease=0.0, min_impurity_split=None,
                      min_samples_leaf=1, min_samples_split=2,
                       min_weight_fraction_leaf=0.0, n_estimators=10,
                       n_jobs=None, oob_score=False, random_state=0, verbose=0,
                       warm start=False)
0.9981965734896303
[[ 909 2]
 [ 4 2412]]
Process finished with exit code 0
```

```
COMPUTING HOG for Negative Files of Training Dataset
HOG computation completed!
(3327, 3780)
(3327,)
Start Learning SVM.
Accuracy: 1.0
finish learning SVM.
Fitness LinearSVC(C=1.0, class_weight=None, dual=True, fit_intercept=True,
          intercept_scaling=1, loss='squared_hinge', max_iter=1000,
          multi_class='ovr', penalty='12', random_state=None, tol=0.0001,
          verbose=0)
Score 1.0
Confusion Matrix showing TP, TN, FP, FN of SVM [[ 911 0]
 0 241611
Training Random Forest
C:\Users\eshas\AppData\Roaming\Python\Python\Python36\site-packages\sklearn\ensemble\forest.py:245: FutureWarning: The default value of n_estimators will change from 10 "10 in version 0.20 to 100 in 0.22.", FutureWarning)
Training Complete
Accuracy: 0.9981965734896303
Finish learning Random Forest.
Random Forest Fitness RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                        max_depth=None, max_features='auto', max_leaf_nodes=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min_samples_leaf=1, min_samples_split=2,
                        min_weight_fraction_leaf=0.0, n_estimators=10,
                        n_jobs=None, oob_score=False, random_state=0, verbose=0,
                        warm_start=False)
Random Forest Score 0.9981965734896303
Confusion Matrix showing TP, TN, FP, FN of RF [[ 909 2]
 [ 4 2412]]
```

COMPUTING HOG for Positive Files of Training Dataset

F1-Score [0.99671053 0.99875776]