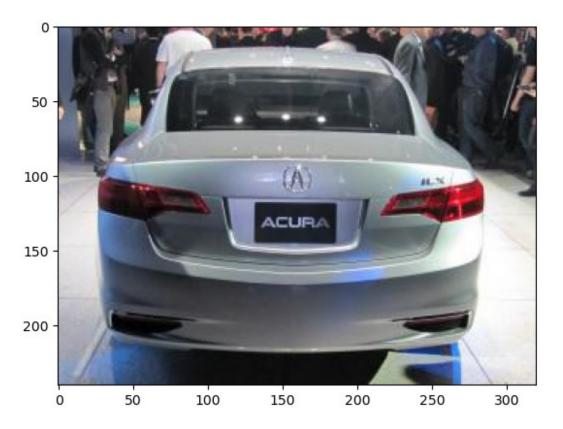
```
import numpy as np
import pandas as pd
import os
import time
import random
import matplotlib.pyplot as plt
import matplotlib.image as mpimg
import seaborn as sns
from glob import glob
from sklearn.svm import SVC
from sklearn.metrics import accuracy score, confusion matrix
from sklearn.model selection import train test split
from sklearn.model selection import GridSearchCV
from skimage.feature import hog
import PIL
import cv2 #must install using "conda install -c conda-forge opency"
into the terminal
print(cv2. version )
import pickle
4.10.0
car paths = glob("../input/the-car-connection-picture-dataset"+"/*")
[:5000] #must download https://www.kaggle.com/datasets/prondeau/the-
car-connection-picture-dataset?resource=download
neg paths = []
for class path in glob("../input/natural-images/natural images"+"/*"):
    if class_path != "../input/natural-images/natural_images/car":
#must dowload https://www.kaggle.com/datasets/prasunroy/natural-
imagesv
        paths = random.choices(glob(class path+"/*"), k=700)
        neg paths = paths + neg paths
print("There are {} car images in the dataset".format(len(car paths)))
print("There are {} negative images in the
dataset".format(len(neg paths)))
There are 5000 car images in the dataset
There are 5600 negative images in the dataset
def process(filename: str=None) -> None:
    View multiple images stored in files, stacking vertically
    Arguments:
        filename: str - path to filename containing image
    image = mpimg.imread(filename)
```

```
# <something gets done here>
plt.figure()
plt.imshow(image)

for file in car_paths[:1]:
    process(file)
```



```
for file in neg_paths[1020:1022]:
    process(file)

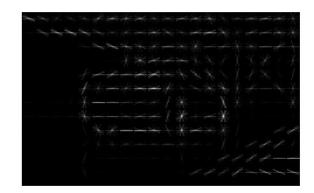
example_image = np.asarray(PIL.Image.open(car_paths[51]))
fig = plt.figure(figsize=(12,6))
plt.imshow(example_image)

<matplotlib.image.AxesImage at 0x1cd953797f0>
```



```
example_image.shape
(200, 320, 3)
from skimage.feature import hog
hog features, visualized = hog(image=example image,
                              orientations=9,
                              pixels_per_cell=(16,16),
                              cells_per_block=(2,2),
                              visualize=True,
                              channel_axis=-1)
fig = plt.figure(figsize=(12,6))
fig.add_subplot(1,2,1)
plt.imshow(example_image)
plt.axis("off")
fig.add_subplot(1,2,2)
plt.imshow(visualized,cmap="gray")
plt.axis("off")
plt.show()
```



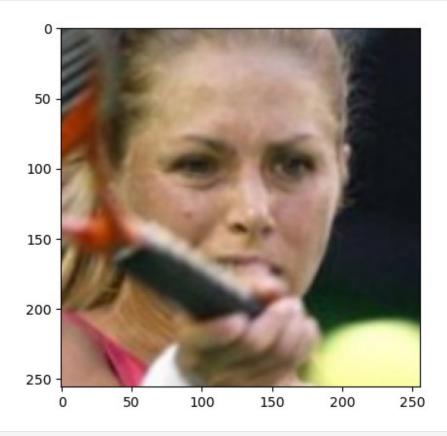


```
hog features.shape
(7524,)
pos images = []
neg images = []
pos labels = np.ones(len(car paths))
neg labels = np.zeros(len(neg paths))
start = time.time()
for car path in car paths:
    img = np.asarray(PIL.Image.open(car path))
    # We don't have to use RGB channels to extract features, Grayscale
is enough.
    img = cv2.cvtColor(cv2.resize(img, (96,64)),cv2.COLOR RGB2GRAY)
    img = hog(img, orientations = 9, pixels per cell = (16, 16),
              cells_per_block=(2,2)
    pos images.append(img)
for neg path in neg paths:
    img = np.asarray(PIL.Image.open(neg path))
    img = cv2.cvtColor(cv2.resize(img, (96,64)), cv2.COLOR RGB2GRAY)
    img = hog(img,orientations=9,pixels_per_cell=(16,16),
              cells per block=(2,2)
    neg images.append(img)
x = np.asarray(pos_images + neg_images)
y = np.asarray(list(pos labels) + list(neg labels))
processTime = round(time.time()-start,2)
print("Reading images and extracting features has taken {}
seconds".format(processTime))
```

```
print("Shape of image set",x.shape)
print("Shape of labels", y.shape)
Reading images and extracting features has taken 17.44 seconds
Shape of image set (10600, 540)
Shape of labels (10600,)
x train,x test,y train,y test =
train_test_split(x,y,test_size=0.2,random state=42)
print(x train.shape)
print(x test.shape)
print(y_train.shape)
print(y test.shape)
(8480, 540)
(2120, 540)
(8480,)
(2120,)
from sklearn.svm import SVC
# Creating a SVC object
svc = SVC(kernel = 'rbf', verbose=True, random state=42)
# Fit the training dataset
svc.fit(x_train,y_train)
y pred = svc.predict(x test)
print("Accuracy score of model is
",round(accuracy score(y pred=y pred,y true=y test)*100, 2))
[LibSVM]Accuracy score of model is 96.51
# testing
def test prediction(img path, true label, svc):
    img = np.asarray(PIL.Image.open(img path))
    img gray = cv2.cvtColor(cv2.resize(img,
(96,64)),cv2.COLOR RGB2GRAY)
    image, viz = hog(img gray, orientations = 9, pixels per cell = (16, 16),
              cells per block=(2,2), visualize=True)
    x tst = np.asarray(image)
    pred = svc.predict([x tst])
    print("True label: ", true_label)
    plt.imshow(img)
    if pred[0] == 0.0:
        print("Prediction: Not car")
    elif pred[0] == 1.0:
        print("Prediction: Car")
```

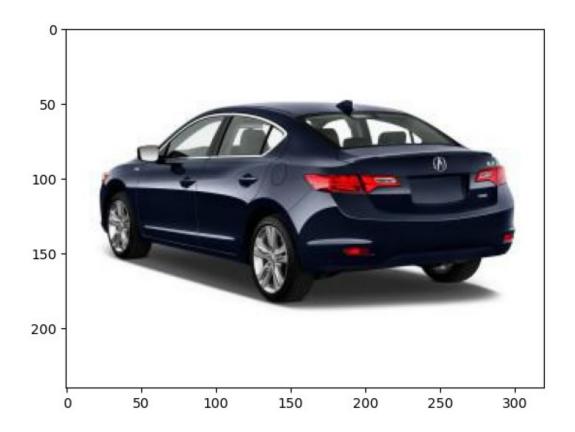
test\_prediction(neg\_paths[2], "Not car", svc)

True label: Not car Prediction: Not car



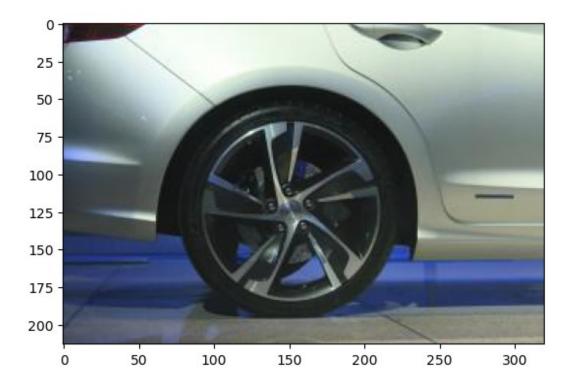
test\_prediction(car\_paths[15], "Car", svc)

True label: Car Prediction: Car



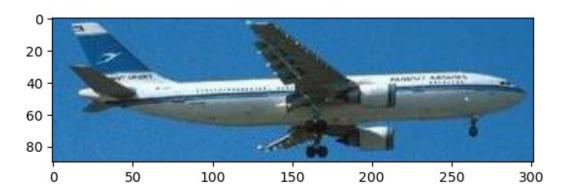
test\_prediction(car\_paths[3], "Car", svc)

True label: Car Prediction: Car



test\_prediction(neg\_paths[-1], "Not car", svc)

True label: Not car Prediction: Not car



```
# Creating a SVC object
svc = SVC()
svc1 = SVC(C=0.01)
svc2 = SVC(C=100)

# Fitting the parameters
svc.fit(x_train,y_train)
svc1.fit(x_train,y_train)
svc2.fit(x_train,y_train)
```

```
y_pred = svc.predict(x_test)
print("Accuracy score of svc model is
",round(accuracy_score(y_pred=y_pred,y_true=y_test)*100,2))

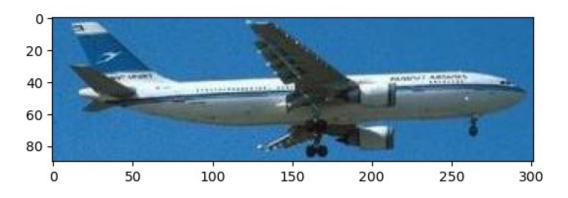
y_pred = svcl.predict(x_test)
print("Accuracy score of svcl model with c=0.01 is
",round(accuracy_score(y_pred=y_pred,y_true=y_test)*100,2))

y_pred = svc2.predict(x_test)
print("Accuracy score of svc2 model with c=100 is
",round(accuracy_score(y_pred=y_pred,y_true=y_test)*100,2))

Accuracy score of svc model is 96.51
Accuracy score of svc1 model with c=0.01 is 88.58
Accuracy score of svc2 model with c=100 is 97.03

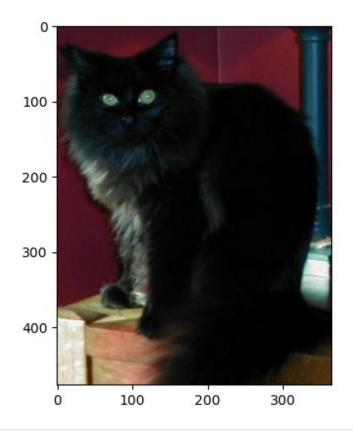
test_prediction(neg_paths[-1], "Not car", svc2)

True label: Not car
Prediction: Not car
```



test prediction(neg paths[4000], "Not car", svc2)

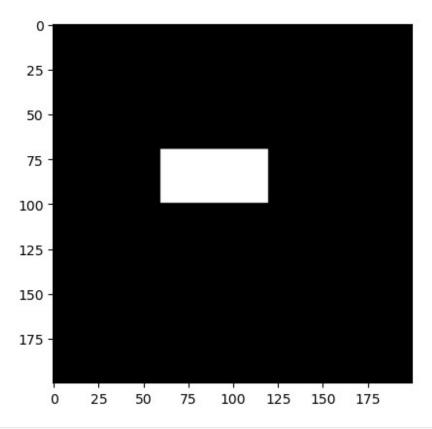
True label: Not car Prediction: Not car



```
from sklearn.model selection import cross val score
scores = cross_val_score(svc, x_test, y_test, cv=5)
scores1 = cross_val_score(svc1, x_test, y_test, cv=5)
scores2 = cross_val_score(svc2, x_test, y_test, cv=5)
print("Accuracy for svc with c=1.0: %0.2f (+/- %0.2f)" %
(scores.mean(), scores.std() * 2))
print("Accuracy for svc with c=0.01: %0.2f (+/- %0.2f)" %
(scores1.mean(), scores1.std() * 2))
print("Accuracy for svc with c=100: %0.2f (+/- %0.2f)" %
(scores2.mean(), scores2.std() * 2))
Accuracy for svc with c=1.0: 0.94 (+/- 0.03)
Accuracy for svc with c=0.01: 0.82 (+/- 0.03)
Accuracy for svc with c=100: 0.94 (+/- 0.02)
def slideExtract(image, windowSize=(96,64), channel="RGB", step=12):
    # Converting to grayscale
    if channel == "RGB":
        img = cv2.cvtColor(image,cv2.COLOR RGB2GRAY)
    elif channel == "BGR":
        img = cv2.cvtColor(image,cv2.COLOR BGR2GRAY)
    elif channel.lower()!="grayscale" or channel.lower()!="gray":
        raise Exception("Invalid channel type")
```

```
# We'll store coords and features in these lists
    coords = []
    features = []
    hIm,wIm = image.shape[:2]
    # W1 will start from 0 to end of image - window size
    # W2 will start from window size to end of image
    # We'll use step (stride) like convolution kernels.
    for w1,w2 in zip(range(0,wIm-
windowSize[0], step), range(windowSize[0], wIm, step)):
        for h1,h2 in zip(range(0,hIm-
windowSize[1],step),range(windowSize[1],hIm,step)):
            window = img[h1:h2,w1:w2]
            features of window =
hog(window, orientations=9, pixels per cell=(16, 16),
                                     cells per block=(2,2)
            coords.append((w1,w2,h1,h2))
            features.append(features of window)
    return (coords,np.asarray(features))
example image = np.asarray(PIL.Image.open("../input/the-car-
connection-picture-dataset/
Acura ILX 2013 28 16 110 15 4 70 55 179 39 FWD 5 4 4dr Mro.jpg"))
coords,features = slideExtract(example image,channel="RGB")
coords[:5]
[(0, 96, 0, 64),
(0, 96, 12, 76),
 (0, 96, 24, 88),
 (0, 96, 36, 100),
 (0, 96, 48, 112)]
from sklearn.preprocessing import MinMaxScaler
class Heatmap():
    def init (self, original image):
        # Mask attribute is the heatmap initialized with zeros
        self.mask = np.zeros(original image.shape[:2])
    # Increase value of region function will add some heat to heatmap
```

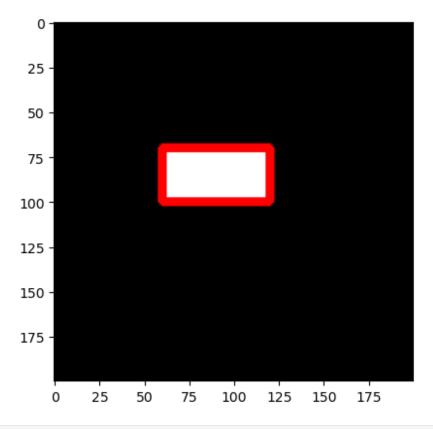
```
def incValOfReg(self,coords):
        w1, w2, h1, h2 = coords
        self.mask[h1:h2,w1:w2] = self.mask[h1:h2,w1:w2] + 30
    # Decrease value of region function will remove some heat from
heatmap
    # We'll use this function if a region considered negative
    def decValOfReg(self,coords):
        w1, w2, h1, h2 = coords
        self.mask[h1:h2,w1:w2] = self.mask[h1:h2,w1:w2] - 30
    def compileHeatmap(self):
        # As you know, pixel values must be between 0 and 255 (uint8)
        # Now we'll scale our values between 0 and 255 and convert it
to uint8
        # Scaling between 0 and 1
        scaler = MinMaxScaler()
        self.mask = scaler.fit transform(self.mask)
        # Scaling between 0 and 255
        self.mask = np.asarray(self.mask * 255).astype(np.uint8)
        # Now we'll threshold our mask, if a value is higher than 170,
it will be white else
        # it will be black
        self.mask = cv2.inRange(self.mask, 170, 255)
        return self.mask
example mask = np.zeros((200,200))
example mask[70:100,60:120] = 255
plt.imshow(example mask,cmap="gray")
plt.show()
```



```
# Find contours function of openCV will help us the find white
regions.
contours, hierarchy =
cv2.findContours(example_mask.astype(np.uint8),1,2)[-2:]
for c in contours:
    if cv2.contourArea(c) < 10*10:
        continue
    (x,y,w,h) = cv2.boundingRect(c)

    rgb_ver =
cv2.cvtColor(example_mask.astype(np.uint8),cv2.CoLOR_GRAY2RGB)
    im = cv2.rectangle(rgb_ver,(x,y),(x+w,y+h),(255,0,0),3)

plt.imshow(im)
<matplotlib.image.AxesImage at 0x1cd9a227140>
```



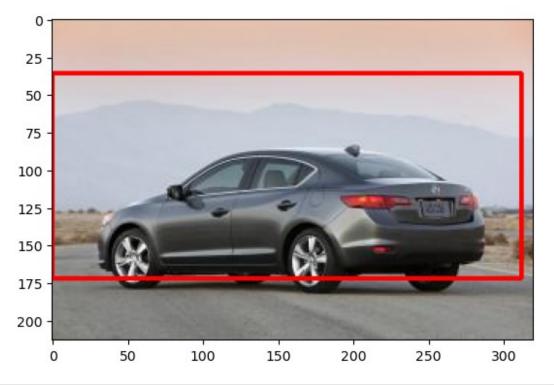
```
def detect(image):
    # Extracting features and initalizing heatmap
    coords,features = slideExtract(image)
    htmp = Heatmap(image)
    for i in range(len(features)):
        # If region is positive then add some heat
        decision = svc.predict([features[i]])
        if decision[0] == 1:
            htmp.incValOfReg(coords[i])
            # Else remove some heat
        else:
            htmp.decValOfReg(coords[i])
    # Compiling heatmap
    mask = htmp.compileHeatmap()
    cont,_ = cv2.findContours(mask,1,2)[:2]
    for c in cont:
        # If a contour is small don't consider it
        if cv2.contourArea(c) < 70*70:
            continue
```

```
(x,y,w,h) = cv2.boundingRect(c)
image = cv2.rectangle(image,(x,y),(x+w,y+h),(255),2)

return image

detected = detect(np.array(PIL.Image.open("../input/the-car-
connection-picture-dataset/
Acura_ILX_2013_28_16_110_15_4_70_55_179_39_FWD_5_4_4dr_ylA.jpg")).copy
())
plt.imshow(detected)

<matplotlib.image.AxesImage at 0x1cd9a25fb30>
```



plt.axis("off")
plt.show()

