# Project 3 Writeup

#### **Instructions**

- Provide an overview about how your project functions.
- Describe any interesting decisions you made to write your algorithm.
- Show and discuss the results of your algorithm.
- Feel free to include code snippets, images, and equations.
- List any extra credit implementation and result (optional).
- Use as many pages as you need, but err on the short side.
- · Please make this document anonymous.

## **Project Overview**

This project implements scene recognition with 3 methods:

- tiny image representation and nearest neighbor classifier.
- bag of words representation and nearest neighbor classifier
- bag of words representation and linear SVM classifier

## **Implementation Detail**

I first implemented get\_tiny\_images. This function was straightforward to code by following the code template instructions. For each image path given in the input I loaded the image, scaled it to 16x16 pixels, reshaped it into a 1-D array, and appended all these features for each image together in a resulting array.

I then implemented nearest\_neighbor\_classify. After getting the distances between the test image features and the training image features, I sorted the distances using np.argsort and grabbed the first k elements of each row, resulting in a 2-D array where each row has the k indices of the closest distances for each image. The labels were then grabbed for each image and I used stats.mode to get the mode for each image.

Next was to implement build\_vocabulary. I used skimage.feature.hog to get the HOG features of each image (after scaling them to 160x160 px). I used 4x4 pixels

per cell and 2x2 cells per block. Then I reshaped the descriptor into a list of 2\*2\*9 block feature vectors and added these features to the bag of words. After doing this for every image, I ran MiniBatchKMeans on the bag of words (setting max\_iter to 200 and n\_clusters to vocab\_size) and returned the cluster centers.

For get\_bag\_of\_words, for each image I similarly got the HOG features using the same settings as in build\_vocabulary. Then I calculated the euclidean distances between the hog features and the input vocabulary. I used np.argsort to grab the index of the smallest distance for each feature and for each smallest distance I added 1 to that bin in the histogram.

Finally for  $svm_classify\ I$  used sklearn.svm. Linear SVC to fit the training image features and returned the predictions on the test features.

#### Result

- 1. Tiny image representation and nearest neighbor classifier accuracy: 20.5%
- 2. Bag of words representation and nearest neighbor classifier accuracy: 53.0%
- 3. Bag of words representation and linear SVM classifier accuracy: 50.1%

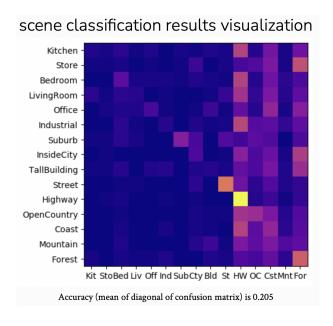
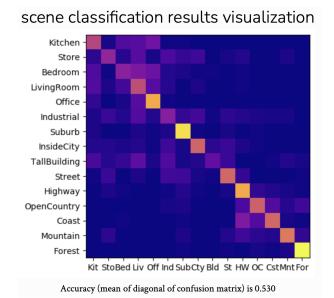


Figure 1: Tiny image and Nearest Neighbor



#### Figure 2: Bag of Words and Nearest Neighbor

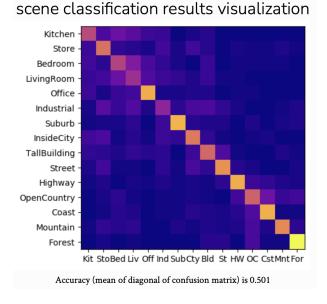


Figure 3: Bag of Words and Linear SVM

# Extra Credit (Optional)

1. I trained the SVM with RBF, which increased the bag of words x svm accuracy to 61.9% The only thing I changed was:

```
# linear classifier:
# lin_clf = LinearSVC()
# updating it to rbf classifier instead:
lin_clf = SVC(kernel='rbf')
```

## scene classification results visualization Bedroom LivingRoom Office Industrial Suburb InsideCity TallBuilding Street Highway OpenCountry Coast Mountain Forest Kit StoBed Liv Off Ind SubCty Bld St HW OC CstMntFor Accuracy (mean of diagonal of confusion matrix) is 0.619

Figure 4: Bag of Words and RBF