Name: SHAH, MIT KALPESHKUMAR UT EID: mks3226

Visual Recognition

#### Coding Assignment 2- Recognizing scene categories

## Implementation:

AlexNet implementation from Caffe was used.

In particular, I used it for 4 different cases:

Case A: No pre-training, No extra-images

**Case B:** With pre-training, No extra-images

**Case C:** No pre-training, With extra-images

Case D: With pre-training, With extra-images

# **Pre-processing**

- First of all, training.txt & testing.txt files were generated from the given 'filenames.mat'.
- Then they were converted to "Imdb format" by running the script /caffe/examples/imagenet/create imagenet.sh
- After that 'scene\_mean.binaryproto' file (mean image file) was created using /caffe/examples/imagenet/make\_imagenet\_mean.sh
- Then, train\_val.protxt & solver.protxt were copied from /caffe/models/bvlc\_Alexnet and path changes were made accordingly.
- fc8 in train\_val.protxt was modified to be of 25 instead of 1000 for the current task.
- Specific changes in parameters (mainly in solver.protxt) are mentioned for each of the cases separately.

Details for each of the case are as follows:

## <u>Case A: No pre-training, No extra-images</u>

Only 2500 training images mentioned were used for it & was tested on 625 images.

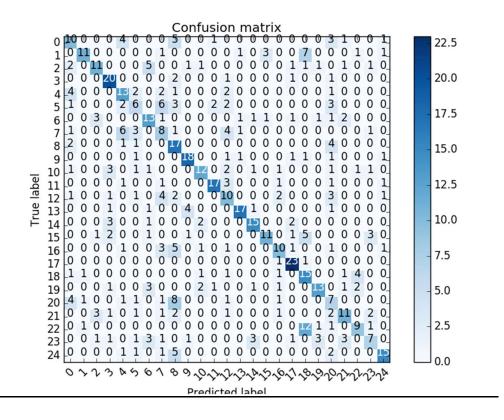
## **Training**

- Network was initialized randomly
- Training was done for two learning rates 0.01, 0.001

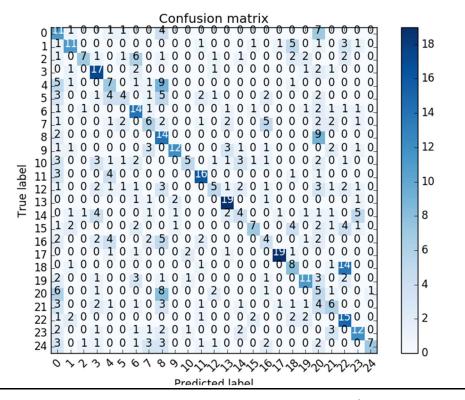
#### Logs

- Examining the logs, maximum accuracy on the test set was:
  - o LR=0.01 => **41.92%** after **4400** iterations
  - o LR=0.001 => **56.16**% after **10200** iterations
- For learning rate **0.01**, after 4500 iterations, accuracy saturated and did not increase further. It kept floating somewhere between 39% & 41% & after 6000 iterations, it started decreasing on a long term and reached till a low of 16%.
- For learning rate **0.001**, accuracy was steadily non-decreasing till around 10k iterations, and after it reached a final maximum of 56.16%.

- Confusion Matrix & Analysis
  - Accuracy from confusion matrix for both the learning rates are 39.54% & 51.28%. (It's lower than one in the logs because the snapshot was saved at 4000, 10000 rather than 4400, 10200 iterations.)
  - From both the confusion matrix it looks like classes Street(4) & Sand(18) performed the best; while Observatory(24), ShoeShop(21) & DormRoom(6) performed worst.
  - It can be seen that one with LR=0.001, is more concentrated on the diagonal.



## CASE A, LR = 0.001, 51.28%



CASE A, LR = 0.01, 39.54%

# <u>Case B: With</u> pre-training, <u>No</u> extra-images

Only 2500 training images mentioned were used for it & was tested on 625 images.

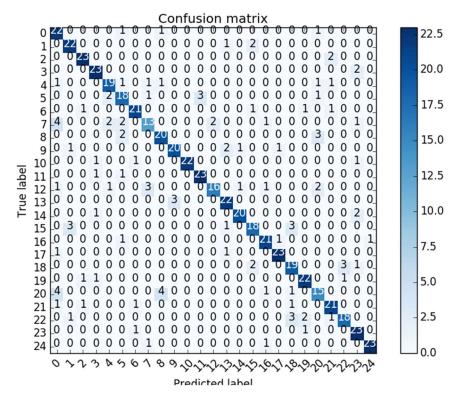
## **Training**

- Network was initialized with pre-trained weights of ImageNet using /caffe/models/bvlc\_reference\_caffenet/ bvlc\_reference\_caffenet.caffemodel
- FineTuning was done with learning rate 0.001
- For the last layer **fc8**, **LR Multipliers** were kept at **10**, **20**; unlike other layers where they were 1,2

### **Logs**

- Examining the logs, maximum accuracy on the test set was:
  - o **82.72%** after **2800** iterations
- After the first 200 iterations only, accuracy was at 78%. Over next 5000 iterations, it kept floating between 76% & 83%, with maximum of 82.72% at 2800 iterations

- Confusion Matrix & Analysis
  - Accuracy on confusion matrix is 81.51%
  - Classes Hospital(8), Operating(13), ShoeShop(21), does not seem to perform well; while almost all are performing very well.



**CASE B, 81.51%** 

# <u>Case C: No pre-training, With extra-images</u>

Other than given 2500 training images, extra 4105 images from the SUN dataset only were used for training, thus a total of 6605 images & was tested on 625 images.

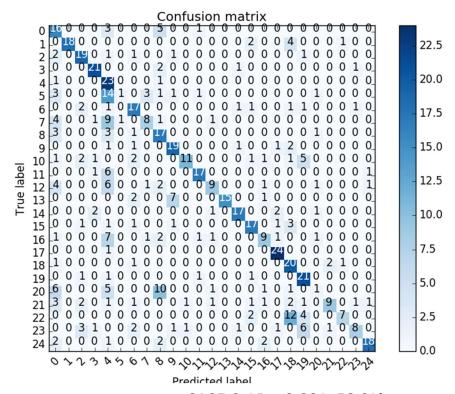
# <u>Training</u>

- Network was initialized randomly
- Training was done for **two learning rates** 0.01, 0.001

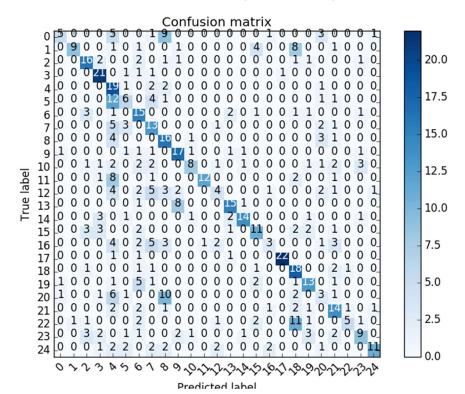
#### Logs

- Examining the logs, maximum accuracy on the test set was:
  - o LR=0.001 => **64.48%** after **8400** iterations
  - o LR=0.01 => **53.28**% after **5600** iterations
- Overall behavior for both learning rates was similar as in case A.

- Confusion Matrix & Analysis
  - Accuracy on confusion matrix are 58.2% & 48.07%
  - Hospital(8), Railway(11), Operating(13), Garrage(17),
    ShoeShop(21), FormalGarden(23) are performing worst, while Dining Room(5), Sand(18) are performing best.
  - Here also, as in class A, it can be seen that confusion matrix with LR=0.001 is more concentrated on diagonal.



CASE C, LR = 0.001, 58.2%



**CASE C, LR = 0.01, 48.07%** 

# Case D: With pre-training, With extra-images

Other than given 2500 training images, extra 4105 images from the SUN dataset only were used for training, thus a total of 6605 images & was tested on 625 images.

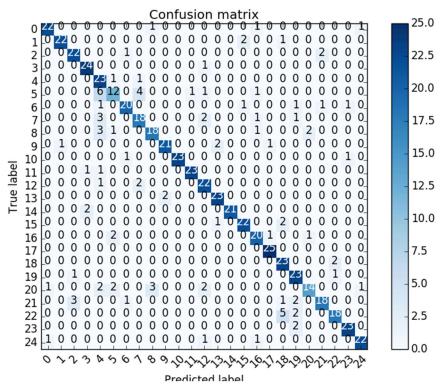
## **Training**

- Network was initialized with pre-trained weights of ImageNet using /caffe/models/bvlc\_reference\_caffenet/ bvlc\_reference\_caffenet.caffemodel
- FineTuning was done with learning rate 0.001
- For the last layer **fc8**, **LR Multipliers** were kept at **10**, **20**; unlike other layers where they were 1,2

#### Logs

- Examining the logs, maximum accuracy on the test set was:
  - o **84.96%** after **10000** iterations
- Again as in case B, because of pretraining, it reached to accuracy of around 78% after 200 iterations only. At 10000 iterations, it reached a maximum of 84.96.

- Confusion Matrix & Analysis
  - O Accuracy on confusion matrix is 83.9%
  - Here almost every class is performing very well, compared to previous 3 cases, & everything is concentrated on diagonal.



**CASE D, 83.9%** 

## **Conclusion:**

From these 4 cases, it can be concluded that

- PreTraining with ImageNet <u>HELPS A LOT</u> in increasing the accuracy.!
- Also, training with **extra Images** do help <u>up to some level</u> in increasing accuracy.
- While learning from scratch, **learning rate** can have a <u>major impact</u> on final accuracy, as can be seen in cases A & C, with learning rates 0.01 & 0.001.
- Classes such as ShoeShop, Operating, Observatory are performing worse consistently, while class Sand is performing best.