## **Relative-Rotations**

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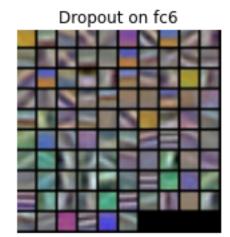
April 3, 2015

Things I will try the PASCAL/Imagenet Setting: 1. Take bounding boxes that are just rescaled to constant 256 x 256 2. Try cropping the boxes carefully. 3. On Picking run the experiment of trying with different layers - how well the network generalizes. Create a pose embedding. Do visual reasoning in this domain. ie. two buses subtracted from each other + car shifts the car in the right way. Initially I was making an error that mirror flipping was on while predicting the relative rotations. Even with this I got some reasonable weights while training from scratch. They are plotted below.

```
%matplotlib inline
          %load_ext autoreload
In [19]:
          %autoreload 2
                      = '/work4/pulkitag-code/pkgs/caffe-v2-2/modelFiles/keypoints/code/'
= '/work4/pulkitag-code/pkgs/caffe-v2-2/modelFiles/keypoints/exp/'
         codeDir
          snapshotDir = '/data1/pulkitag/snapshots/keypoints/'
          import caffe
          import os
          import my_pycaffe as mp
          import matplotlib.pyplot as plt
          #import plotly.plotly as ply
          #plotly.tools.set_credentials_file(username='pulkit.audacious', api_key='9cqniif4ai',
          #Experiment and snapshot paths.
          #Load the module
          currDir = os.getcwd()
         os.chdir(codeDir)
          import process3d as p3d
         os.chdir(currDir)
         expStr = 'rotObjs_128_kmedoids30_20'
          defFile = os.path.join(expDir, expStr, 'caffenet_siamese.prototxt')
         The autoreload extension is already loaded. To reload it, use:
            %reload_ext autoreload
              #This network was trained by concatenating fc-6 features. Prior to concatenation t
              numIterations = 60000
In [13]:
              numIterations2 = 90000
                             = 128
              imSz
              expStr
                       = 'rotObjs_128_kmedoids30_20'
              modelName = 'keypoints_siamese_scratch_iter_%d.caffemodel' % numIterations
              defFile = os.path.join(expDir, expStr, 'keynet_siamese_deploy.prototxt')
              modelFile = os.path.join(snapshotDir, 'exprotObjs_lblkmedoids30_20_im$z%d'% imSz,
                        = mp.MyNet(defFile, modelFile)
              #Network without dropouts on fc-6 features.
              modelName2 = 'keypoints_siamese_scratch_nodrop_fc6_iter_%d.caffemodel' % numIterat
              modelFile2 = os.path.join(snapshotDir,"exprotObjs_lblkmedoids30_20_imSz%d" % imSz
                          = mp.MyNet(defFile, modelFile2)
              fig = plt.figure()
              ax1 = plt.subplot(1,2,1)
              ax2 = \overline{plt.subplot(1,2,2)}
              net.vis_weights('conv1', ax=ax1, titleName='Dropout on fc6')
net2.vis_weights('conv1', ax=ax2, titleName='No Dropout on fc6')
              ax1.axis('off')
              ax2.axis('off')
```

(-0.5, 79.5, 79.5, -0.5)

Out [13]:

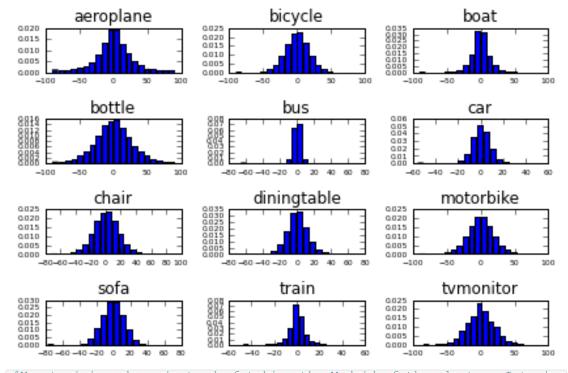


## No Dropout on fc6

```
In []:
       #Define the experiment
In [17]: prms = p3d.get_exp_prms(imSz=128, lblType='uniform', numSamplesTrain=40000, numSamples
                               azBin=30, elBin=10)
         #Get the statistics of the training data.
        #Training data is PASCAL 2012 + Entire Imagenet.
        annCountTrain = p3d.get_class_statistics_exp(prms, 'train')
        #Get the statistics of the val data
        annCountVal = p3d.get_class_statistics_exp(prms, 'val')
        annCount = {'%s' % key: [val] for key, val in annCountTrain.iteritems()}
In [18]: annCount = {'%s' % key: annCount[key] + [val] for key, val in annCountVal.iteritems()}
        formatStr = "{:<12} {:<20} "
        print formatStr.format('Class','Train-BBox', 'Val-BBox')
        print '-' * 100
        for key, val in sorted(annCount.items()):
            bbTr, bbVl = val
            print ("%s" % formatStr).format(key, bbTr, bbVl)
               Train-BBox
        Class
                                          Val-BBox
        _____
        aeroplane 2314
                                          361
                   1656
                                          340
        bicycle
                     2489
                                          265
        boat
        bottle
                    1696
                                          272
                    1356
                                          283
        bus
        car
                     6194
                                          594
                                          1032
                    2059
        chair
        diningtable 2611
                                          288
                    1570
                                          339
        motorbike
        sofa
                     1749
                                          294
                    1601
                                          298
        train
        tymonitor 1570
                                          336
        # Get statistics of pairwise labels
In [6]: lblType = 'diff'
        numSamples = 20000
        imSz = 128
cropType = 'contPad'
```

```
= p3d.get_exp_prms(imSz=imSz, lblType=lblType, numSamples=numSamples, cropT
           pairLabels = p3d.get_pair_label_stats(prms, setName='train')
           count = 1
           for (key, val) in sorted(pairLabels.items()):
In [19]:
                         = plt.subplot(4,3,count)
                ax
                azimuth = val[:,0]
                n, bins, patches = plt.hist(azimuth, 20, normed=1, histtype='bar', rwidth=0.8)
                plt.title(key)
                ax.tick_params(axis='x', labelsize=6)
ax.tick_params(axis='y', labelsize=6)
                count += 1
           plt.tight_layout()
                      aeroplane
                                                      bicycle
                                                                                     boat
            00035
00035
                 209-159-100-50 0 50 100 150 200
                                               209-159-100-50 0 50 100 150 200
                                                                             200150-100-50 0 50 100 150 200
                        bottle
                                                        bus
                                                                                      car
             0.05
              0.04
              0.03
              0.02
              0.01
              0.00
                -150-100-50 0 50 100 150 200
                                              -209-159-100-50 O 50 100 150 200
                                                                            -20#159-100-50 O 50 100 150 200
                         chair
                                                   diningtable
                                                                                  motorbike
             0.005
             0.004
             0.003
             0.001
             0.000
                -209-150-100-50 O 50 100 150 200
                                              -200-159-100-50 O 50 100 150 200
                                                                            -20<del>9</del>159-100-50 0 50 100 150 200
                          sofa
                                                       train
                                                                                  tymonitor
             0.005
             0.006
             0.002
                                                            50 100 150 200
                -209-159-100-50 O 50 100 150 200
                                              -209-159-100-50 O
                                                                            -209-150-100-50 O 50 100 150 200
          #pairLabels = p3d.get_pair_label_stats(prms, setName='val')
In [30]: | count = 1
           fig = plt.figure()
           for (key, val) in sorted(pairLabels.items()):
                          = plt.subplot(4,3,count)
                azimuth = val[:,1]
                n, bins, patches = plt.hist(azimuth, 20, normed=1, histtype='bar', rwidth=0.8)
                plt.title(key)
                ax.tick_params(axis='x', labelsize=6)
                ax.tick_params(axis='y', labelsize=6)
                count += 1
```

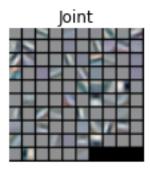
plt.tight\_layout()

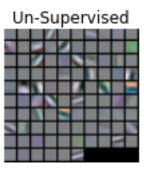


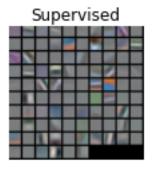
#New training where instead of taking the Medoid of the cluster, I trained on the bins

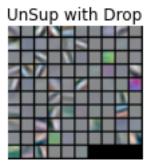
```
# These examples contain information from all boxes whose minimum size > 50px in the o
In [24]:
        prms = p3d.get_exp_prms(cropType='contPad', numSamplesTrain=40000, lblType='uniform')
        numIter = 50000
        fig = plt.figure()
        ax\tilde{1} = plt.subplot(2,2,1)
        ax2 = plt.subplot(2,2,2)
        ax3 = plt.subplot(2,2,3)
        ax4 = plt.subplot(2,2,4)
        ## Get Supervised + rotation loss example
        #Unforutnately this was not done programattically :/
        snapDir = '/data1/pulkitag/pascal3d/snapshots/'
        ("/work4/pulkitag-code/pkgs/caffe-v2-2/modelFiles/pascal3d/exp/"
        defFile =
                     "imSz128_lbl-uni-az30el10_crp-contPad16_ns4e+04_mb50/caffenet_siamese_fc6
        netJoint = mp.MyNet(defFile, modelFile)
        print modelFile
        netJoint.vis_weights('conv1', ax=ax1, titleName='Joint')
        ax1.axis('off')
        #Get the Supervised example.
        cPrms = p3d.get_caffe_prms(isScratch=True, isClassLbl=True, concatLayer='fc6', noRot=T
        exp = p3d.get_experiment_object(prms, cPrms)
        exp.init_from_self()
        modelFile = exp.get_snapshot_name(numIter=numIter)
        defFile
                  = exp.files_['netdef']
                  = mp.MyNet(defFile, modelFile)
        netSup
        print modelFile
        netSup.vis_weights('conv1', ax=ax2, titleName='Supervised')
        ax2.axis('off')
        #Get Unsupervised example.
        cPrms = p3d.get_caffe_prms(isScratch=True, isClassLbl=False, concatLayer='fc6', noRot=
        exp = p3d.get_experiment_object(prms, cPrms)
```

```
exp.init_from_self()
modelFile = exp.get_snapshot_name(numIter=numIter)
defFile = exp.files_['netdef']
netUnsup = mp.MyNet(defFile, modelFile)
print modelFile
netUnsup.vis_weights('conv1', ax=ax3, titleName='Un-Supervised')
ax3.axis('off')
#Get Unsupervised example.
cPrms = p3d.get_caffe_prms(isScratch=True, isClassLbl=False, concatLayer='fc6', noRot=
exp = p3d.get_experiment_object(prms, cPrms)
exp.init_from_self()
modelFile = exp.get_snapshot_name(numIter=numIter)
defFile = exp.files_['netdef']
         = mp.MyNet(defFile, modelFile)
netUnsup
print modelFile
netUnsup.vis_weights('conv1', ax=ax4, titleName='UnSup with Drop')
ax4.axis('off')
plt.tight_layout()
/data1/pulkitag/pascal3d/snapshots/imSz128_lbl-uni-az30el10_crp-
contPad16_ns4e+04_mb50/pascal3d_scratch_concat-
fc6_iter_50000.caffemodel
Ignoring line: # Autotmatically generated solver prototxt
/data1/pulkitag/pascal3d/snapshots/pascal3d_imSz128_lbl-uni-
az30el10 crp-contPad16 ns4e+04 mb50/caffenet scratch sup noRot fc6 ite
r_50000.caffemodel
Ignoring line: # Autotmatically generated solver prototxt
/data1/pulkitag/pascal3d/snapshots/pascal3d_imSz128_lbl-uni-
az30el10_crp-contPad16_ns4e+04_mb50/caffenet_scratch_unsup_fc6_iter_50
000.caffemodel
Ignoring line: # Autotmatically generated solver prototxt
/data1/pulkitag/pascal3d/snapshots/pascal3d_imSz128_lbl-uni-
az30el10_crp-contPad16_ns4e+04_mb50/caffenet_scratch_unsup_fc6_drop_it
er_50000.caffemodel
```









This is weird that the filters with k-Medoid clustering look so much better than filters obtained with the Euler angle uniform binning. The critical differences between the two approaches is as following:

- 1. First only considers small rotations (< 30 degrees) and bins all the other rotation into twenty categories.
- 2. First uses 0.005 weight std to initialize the FC layers, the latter uses 0.01 for initialization. This was done as with 0.005 init, the second case I was getting all zeros as output in the top layers.
- 3. First uses pascal only, whereas second used imagenet also.
- 4. There is a difference in classes that are bein used. First uses a set of 15 manually chosen classes that are mostly rigid, 12 uses the 12 Rigid objects in pascal3d. The first includes the classes of person, sheep, bird, cow and horse that are ignored by the second. The second has bottle and diningtable which is not there in the first.

## Possible things to do:

- 1. Compare accuracy on an auxiliary dataset say Caltech101? Any suggestions?
- 2. The issue with using less data should I use the same architecture as the one used by the unsupervised method which has a lot more training data available? Is that fine?

In []: