In [1]: %matplotlib inline

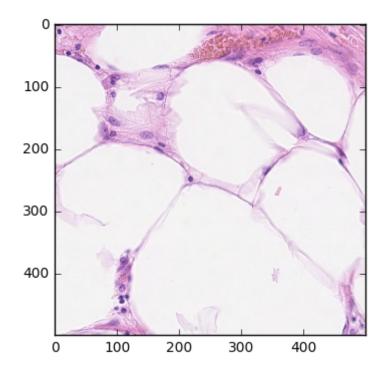
In [118]: import tensorflow as tf import tensorflow.contrib.slim as slim from scipy.io import loadmat import matplotlib import matplotlib.pyplot as plt import matplotlib.image as mpimg import numpy as np import os import itertools import math from ipywidgets import interact, interactive, fixed import ipywidgets as widgets

```
In [351]: from classifier_utils import get_image_data
    from classifier_utils import draw
    from classifier_utils import get_dataset
    from classifier_utils import get_examples
    from classifier_utils import expand_training_data
    from classifier_utils import get_accuracy, get_weighted_f1, get_confusion
```

In [3]: $i = mpimg.imread('Dataset/CRCHistoPhenotypes_2016_04_28/Classification/img1/img1.bmp')$

In [4]: plt.imshow(i)

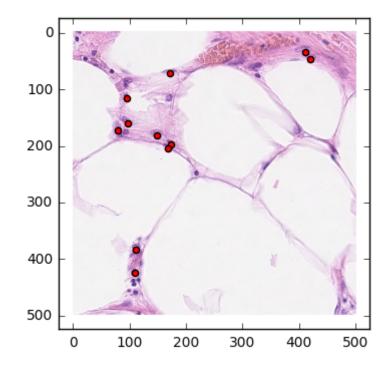
Out[4]: <matplotlib.image.AxesImage at 0x7f15d237e050>



```
d = loadmat('Dataset/CRCHistoPhenotypes 2016 04 28/Classification/img1/im
        g1 fibroblast.mat')['detection']
In [6]:
Out[6]: array([[ 148.78269485,
                                 181.47754293],
                  96.60303831,
                                 159.68097754],
                  79.42998679,
                                 172.89101717],
                                  71.834214
                 171.9002642 ,
                  95.77741083,
                                 116.08784676],
                 410.34147952,
                                  33.19484808],
                [ 419.91875826,
                                  46.73513871],
                 172.89101717,
                                 197.16446499],
                [ 168.2675033 ,
                                 204.42998679],
                [ 110.96895641,
                                 383.42602378],
                [ 108.65719947,
                                 424.70739762]])
```

```
In [7]: plt.imshow(i)
plt.scatter(d[:,0], d[:,1], c='r')
```

Out[7]: <matplotlib.collections.PathCollection at 0x7f15d21fcb90>

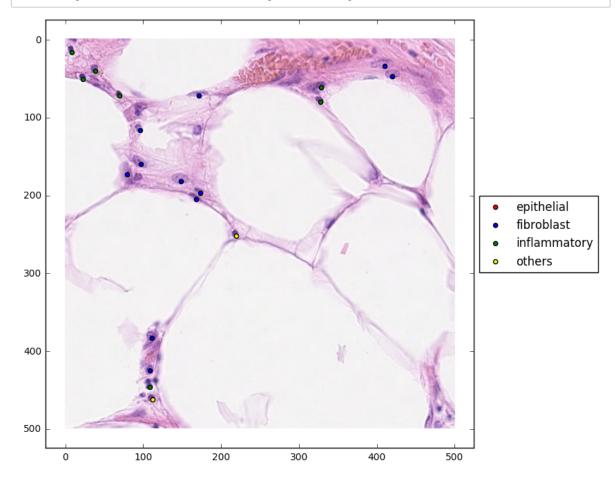


```
In [8]: os.path.splitext("Dataset/CRCHistoPhenotypes_2016_04_28/whatever.bmp")
Out[8]: ('Dataset/CRCHistoPhenotypes_2016_04_28/whatever', '.bmp')
In [9]: categories = [
    'epithelial',
    'fibroblast',
    'inflammatory',
    'others',
```

]

```
(img, centres, labels) = get_image_data('Dataset/CRCHistoPhenotypes_2016_
In [339]:
            04 28/Classification/img1/img1.bmp', categories)
 In [12]:
           np.hstack((centres, labels)).astype('int32')
 Out[12]: array([[148, 181,
                                  0,
                                        1,
                                             0,
                                                   0],
                   [ 96, 159,
                                                   0],
                                  0,
                                        1,
                                             0,
                   [ 79, 172,
                                        1,
                                                   0],
                                  0,
                                             0,
                   [171, 71,
                                  0,
                                        1,
                                             Θ,
                                                   0],
                                                   0],
                   [ 95, 116,
                                             0,
                                  0,
                                        1,
                           33,
                    [410,
                                  0,
                                        1,
                                             0,
                                                   0],
                   [419,
                           46,
                                             0,
                                                   0],
                                  0,
                                        1,
                   [172, 197,
                                             Θ,
                                                   0],
                                  0,
                                        1.
                   [168, 204,
                                        1,
                                             0,
                                                   0],
                                  0,
                   [110, 383,
                                             0,
                                                   0],
                                  0,
                                        1,
                                                   0],
                   [108, 424,
                                  0,
                                        1,
                                             0,
                       8,
                           15,
                                  0,
                                        0,
                                             1,
                                                   0],
                                             1,
                   [ 38,
                           39,
                                                   0],
                                  0,
                                        0,
                   [ 22,
                           50,
                                  0,
                                        Θ,
                                             1,
                                                   0],
                                                   0],
                   [ 69,
                           72,
                                  0,
                                        Θ,
                                             1,
                   [328,
                           60,
                                  0,
                                        0,
                                             1,
                                                   0],
                   [327,
                           79,
                                  0,
                                       0,
                                             1,
                                                   0],
                   [108, 446,
                                             1,
                                  0,
                                        0,
                                                   0],
                   [219, 251,
                                  0,
                                        0,
                                             0,
                                                   1],
                   [111, 462,
                                                   1]], dtype=int32)
                                  0,
                                        0,
                                             0,
```

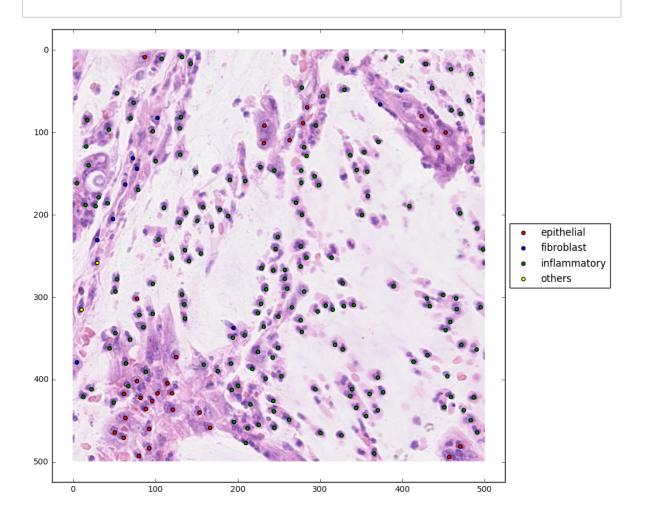
In [340]: draw(img, centres, labels, categories, figsize=(8,8))



In [345]: (i, c, L) = get_dataset(100, categories)

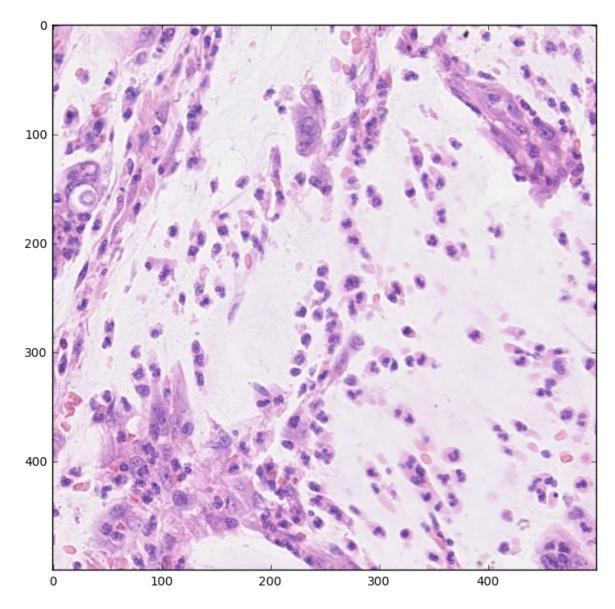
In [18]:

(img, centres, labels) = get_image_data('Dataset/CRCHistoPhenotypes_2016_
04_28/Classification/img6/img6.bmp', categories)
draw(img, centres, labels, categories)



In [19]: plt.figure(figsize=(8,8))
 plt.imshow(img)

Out[19]: <matplotlib.image.AxesImage at 0x7f15cb9b0fd0>



In [22]: patches.shape, labels.shape, centres.shape, img_ids.shape

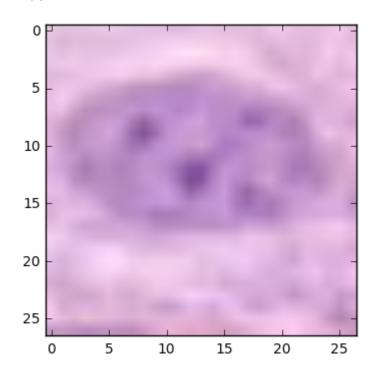
Out[22]: ((20362, 27, 27, 3), (20362, 4), (20362, 2), (20362,))

```
In [24]: test_index = 0
    (x, y) = centres[test_index]
    plt.imshow(i[img_ids[test_index]][y-13:y+14,x-13:x+14,:])
    print np.all(i[img_ids[test_index]][y-13:y+14,x-13:x+14,:] == patches[test_index])
```

True

In [26]:

test labels.shape



```
In [25]:
         patches.shape, labels.shape, centres.shape, img ids.shape
         N = patches.shape[0]
         np.random.seed(0) # predictable shuffling for now
         perm = np.random.permutation(N)
         patches_shuffled = patches[perm]
         labels shuffled = labels[perm]
         centres shuffled = centres[perm]
         img ids shuffled = img ids[perm]
         num train = int(0.8 * N)
         train_patches = patches_shuffled[:num_train]
         train labels = labels shuffled[:num train]
         train centres = centres shuffled[:num train]
         train_img_ids = img_ids_shuffled[:num_train]
         test patches = patches shuffled[num train:]
         test labels = labels shuffled[num train:]
         test centres = centres shuffled[num train:]
         test img ids = img ids shuffled[num train:]
```

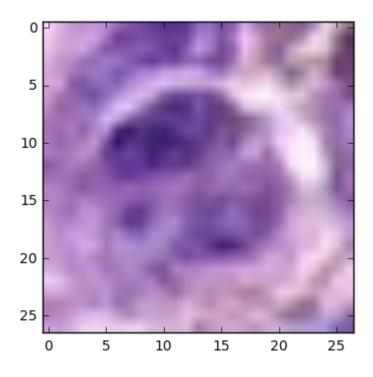
train patches.shape, train labels.shape, test patches.shape,

Out[26]: ((16289, 27, 27, 3), (16289, 4), (4073, 27, 27, 3), (4073, 4))

```
http://localhost:8888/nbconvert/html/Classification%20Experimentation.ipynb?download=false
```

In [27]: plt.imshow(train_patches[0]) print categories[np.argmax(train_labels[0])]

inflammatory



```
In [28]: np.sum(train_labels, axis=0)
```

Out[28]: array([5683, 4074, 5043, 1489])

```
In [277]: try:
    del train_dict # if it exists
    del sorted_train_dict # if it exists
    except:
        pass
    sorted_train_dict = expand_training_data(i, train_patches, train_labels, train_centres, train_img_ids, 15000)
```

In [278]: for (k, v) in sorted_train_dict.iteritems():
 print k, v.shape

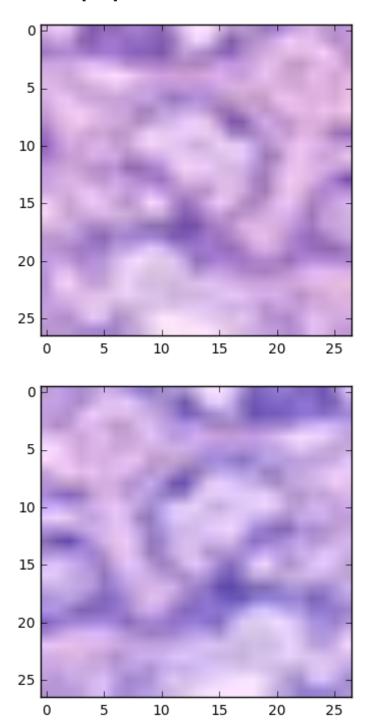
```
hsv_factors (60000, 3)
deltas (60000, 2)
patches (60000, 27, 27, 3)
centres (60000, 2)
img_ids (60000,)
rots (60000,)
labels (60000,)
flips (60000,)
```

```
In [279]: plt.imshow(train_patches[5])
    plt.figure()
    plt.imshow(sorted_train_dict['patches'][1])

print sorted_train_dict['flips'][1], sorted_train_dict['rots'][1], sorted
    _train_dict['deltas'][1]

#print train_patches[4,:3,:3,:]
#print sorted_train_dict['patches'][0,:3,:3,:]
```

True 180 [2 0]



```
In [281]: N = sorted_train_dict['patches'].shape[0]
    np.random.seed(123) # predictable shuffling for now
    perm = np.random.permutation(N)

train_dict = {k : v[perm] for (k, v) in sorted_train_dict.iteritems()}
```

In [283]: sess= None

```
def classifier model(image batch):
In [284]:
              with slim.arg scope([slim.conv2d, slim.fully connected],
                                  activation fn=tf.nn.relu,
                                  weights initializer=tf.truncated normal initializ
          er(stddev=0.01),
                                   biases initializer=tf.zeros initializer,
                                  weights regularizer=slim.l2 regularizer(0.0005),
                                   biases regularizer=None):
                  with slim.arg scope([slim.conv2d, slim.max pool2d], padding='VALI
          D'):
                      with slim.arg scope([slim.dropout], keep prob=0.8):
                          net = image batch
                          net = slim.conv2d(net, 36, [4, 4], scope='1_conv')
                          net = slim.max pool2d(net, [2, 2], scope='2 max pool')
                          net = slim.conv2d(net, 48, [3, 3], scope='3 conv')
                          net = slim.max pool2d(net, [2, 2], scope='4 max pool')
                          net = slim.flatten(net, scope='4 flatten')
                          net = slim.fully connected(net, 512, scope='5 fc')
                          net = slim.dropout(net, scope='5_dropout')
                          net = slim.fully connected(net, 512, scope='6 fc')
                          net = slim.dropout(net, scope='6 dropout')
                          net = slim.fully connected(net, 4, activation fn=None, sc
          ope='7 fc')
                           return net
In [285]: if sess is not None:
              sess.close()
          tf.reset default graph()
          sess = tf.InteractiveSession()
          patch tensor = tf.placeholder(dtype='float32', shape=(None, 27, 27, 3))
          label tensor = tf.placeholder(dtype='int32', shape=(None, 4))
          with tf.variable scope("classifier"):
              with slim.arg scope([slim.dropout], is training=True):
                  prediction logits = classifier model(patch tensor)
          with tf.variable scope("classifier", reuse=True):
              with slim.arg scope([slim.dropout], is training=False):
                  inference prediction logits = classifier model(patch tensor)
          def get_training_op(prediction_logits, labels):
In [286]:
              loss = slim.losses.softmax cross entropy(prediction logits, labels)
              total loss = slim.losses.get total loss()
              optimizer = tf.train.MomentumOptimizer(momentum=0.9, learning rate=5e
          -4)#learning rate=0.01)
              #optimizer = tf.train.AdamOptimizer()
              train op = slim.learning.create train op(total loss, optimizer)
              return (train op, loss)
In [287]:
          (train op, loss) = get training op(prediction logits, label tensor)
```

```
In [288]:
          predictions = slim.softmax(prediction logits)
          inference predictions = slim.softmax(inference prediction logits)
          predictions.get shape(), inference predictions.get shape(),
In [289]:
          label tensor.get shape()
Out[289]: (TensorShape([Dimension(None), Dimension(4)]),
           TensorShape([Dimension(None), Dimension(4)]),
           TensorShape([Dimension(None), Dimension(4)]))
In [293]:
          dropout accuracy = get accuracy(predictions, label tensor)
          accuracy = get accuracy(inference predictions, label tensor)
          f1 = get weighted f1(inference predictions, label tensor)
          confusion = get confusion(inference predictions, label tensor)
          def train loop(sess, patch tensor, label tensor, train patches, train lab
In [311]:
          els, test patches, test labels, train op, softmax loss, epochs, batch siz
          e, predictions, dropout accuracy, accuracy, f1, confusion, reset=True):
              tr loss = []
              tst loss = []
              N = train patches.shape[0]
              assert train labels.shape[0] == N
              if reset:
                  sess.run(tf.initialize all variables())
              for e in xrange(epochs):
                  for i in xrange(0, N, batch size):
                      [_, loss, p] = sess.run([train op, softmax loss,
          predictions], feed dict={
                              patch tensor:train patches[i:i+batch size],
                              label_tensor:train_labels[i:i+batch_size],
                          })
                      step = i / batch_size
                      if step % 25 == 0:
                          [test loss, dacc, acc, f] = sess.run([softmax loss, dropo
          ut accuracy, accuracy, f1], feed dict={
                              patch_tensor:test_patches,
                              label tensor:test labels,
                          })
                          print "Epoch %d, step %d, training loss %f, test loss %f,
           accuracy = f/f, f1 = f'' % (e, step, loss, test loss, dacc, acc, f)
                          tr loss.append(loss)
                          tst loss.append(test loss)
                  [test loss, dacc, acc, f, conf] = sess.run([softmax loss, dropout
          _accuracy, accuracy, f1, confusion], feed_dict={
                      patch tensor:test patches,
                      label tensor:test labels,
                  print "End of epoch %d, training loss %f, test loss %f, accuracy
           = f/f, f1 = f'' % (e, loss, test loss, dacc, acc, f)
                  print "Confusion matrix:"
                  print conf
              return (tr loss, tst loss)
```

In [295]:

no augmentation
#tr_loss, tst_loss = train_loop(sess, patch_tensor, label_tensor, train_p
atches, train_labels, test_patches, test_labels, train_op, loss, 20, 100,

predictions, dropout_accuracy, accuracy, f1, confusion)

with augmentation

tr_loss, tst_loss = train_loop(sess, patch_tensor, label_tensor, train_di
ct['patches'], train_dict['labels'], test_patches, test_labels, train_op,
loss, 10, 100, predictions, dropout accuracy, accuracy, f1, confusion)

```
Epoch 9, step 225, training loss 0.861277, test loss 0.748208, accuracy =
 0.715198/0.722072, f1 = 0.722154
Epoch 9, step 250, training loss 0.857543, test loss 0.749148, accuracy =
 0.714461/0.726737, f1 = 0.729641
Epoch 9, step 275, training loss 0.827899, test loss 0.767274, accuracy =
 0.705868/0.714952, f1 = 0.721258
Epoch 9, step 300, training loss 0.827952, test loss 0.783552, accuracy =
 0.691628/0.700221, f1 = 0.702412
Epoch 9, step 325, training loss 1.056151, test loss 0.782347, accuracy =
 0.697520/0.701694, f1 = 0.705863
Epoch 9, step 350, training loss 0.842201, test_loss 0.756889, accuracy =
 0.711024/0.716916, f1 = 0.722787
Epoch 9, step 375, training loss 0.820612, test loss 0.750400, accuracy =
 0.714461/0.720108, f1 = 0.723351
Epoch 9, step 400, training loss 0.870176, test loss 0.764686, accuracy =
 0.703167/0.710042, f1 = 0.717933
Epoch 9, step 425, training loss 0.905569, test loss 0.699593, accuracy =
 0.731156/0.732875, f1 = 0.723545
Epoch 9, step 450, training loss 0.883726, test loss 0.779491, accuracy =
 0.691137/0.703413, f1 = 0.714516
Epoch 9, step 475, training loss 0.693343, test_loss 0.729867, accuracy =
 0.717653/0.726000, f1 = 0.722386
Epoch 9, step 500, training loss 0.793699, test loss 0.789056, accuracy =
 0.696784/0.705377, f1 = 0.712449
Epoch 9, step 525, training loss 0.998048, test loss 0.763856, accuracy =
 0.706113/0.713725, f1 = 0.716686
Epoch 9, step 550, training loss 0.901893, test loss 0.820946, accuracy =
 0.670513/0.682789, f1 = 0.698408
Epoch 9, step 575, training loss 0.866222, test loss 0.782978, accuracy =
 0.702922/0.711760, f1 = 0.715125
End of epoch 9, training loss 0.828047, test loss 0.724290, accuracy = 0.
731647/0.732138, f1 = 0.732246
Confusion matrix:
[[1083
        170
            105
                   35]
 [ 129
        705
            163
                   721
   31
        100 1035
 Γ
                   871
 Γ
   20
         72
            107
                  159]]
```

In [312]: # Try another 10 epochs tr_loss2, tst_loss2 = train_loop(sess, patch_tensor, label_tensor, train_ dict['patches'], train_dict['labels'], test_patches, test_labels, train_o p, loss, 10, 100, predictions, dropout accuracy, accuracy, f1, confusion, reset=False)

Epoch 9, step 225, training loss 0.665841, test loss 0.743780, accuracy =

```
0.714461/0.719863, f1 = 0.728241
          Epoch 9, step 250, training loss 0.688977, test loss 0.717920, accuracy =
           0.730174/0.744905, f1 = 0.744861
          Epoch 9, step 275, training loss 0.716597, test loss 0.746891, accuracy =
           0.717407/0.729683, f1 = 0.729617
          Epoch 9, step 300, training loss 0.663885, test loss 0.706045, accuracy =
           0.733366/0.741959, f1 = 0.741381
          Epoch 9, step 325, training loss 0.845116, test loss 0.738857, accuracy =
           0.723300/0.736067, f1 = 0.737651
          Epoch 9, step 350, training loss 0.638910, test loss 0.733105, accuracy =
           0.716916/0.724282, f1 = 0.732080
          Epoch 9, step 375, training loss 0.704500, test loss 0.758284, accuracy =
           0.715689/0.722809, f1 = 0.728328
          Epoch 9, step 400, training loss 0.701134, test_loss 0.720230, accuracy =
           0.717898/0.735085, f1 = 0.743507
          Epoch 9, step 425, training loss 0.779204, test loss 0.670482, accuracy =
           0.748588/0.750307, f1 = 0.746413
          Epoch 9, step 450, training loss 0.680333, test loss 0.744098, accuracy =
           0.712497/0.726246, f1 = 0.732868
          Epoch 9, step 475, training loss 0.580466, test_loss 0.695251, accuracy =
           0.743432/0.755463, f1 = 0.756033
          Epoch 9, step 500, training loss 0.601554, test_loss 0.722973, accuracy =
           0.723300/0.732384, f1 = 0.737698
          Epoch 9, step 525, training loss 0.744669, test loss 0.680748, accuracy =
           0.745151/0.750798, f1 = 0.751176
          Epoch 9, step 550, training loss 0.738777, test loss 0.724834, accuracy =
           0.724527/0.741714, f1 = 0.748154
          Epoch 9, step 575, training loss 0.687487, test loss 0.776052, accuracy =
           0.709796/0.718144, f1 = 0.719238
          End of epoch 9, training loss 0.552169, test loss 0.693133, accuracy = 0.
          742941/0.750552, f1 = 0.751103
          Confusion matrix:
          [[1090 206
                        76
                             21]
           「 114   793   112
                             501
              46 130 1002
           ſ
                             75]
              23
                   75
                        88 172]]
           [
          # Save the model
In [313]:
```

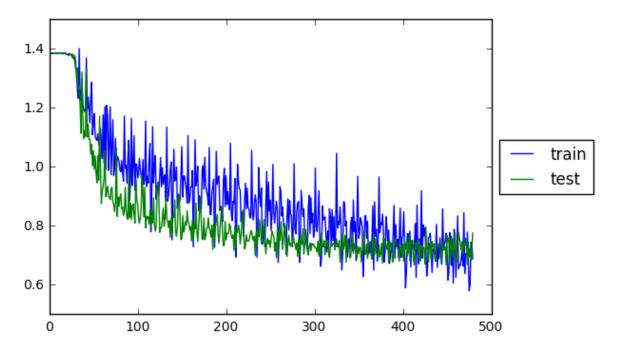
In [313]: # Save the model saver = tf.train.Saver() save_path = saver.save(sess, "classification_models/v3_augmentation/mode l.ckpt") print "Saved to:", save_path

Saved to: classification_models/v3_augmentation/model.ckpt

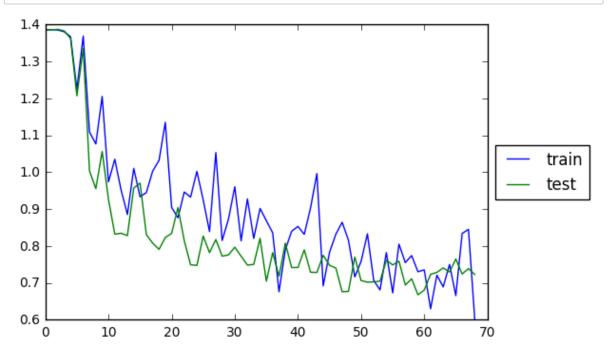
```
In [ ]: # Restore the model
    saver = tf.train.Saver()
    saver.restore(sess, "classification_models/v1/model.ckpt")
```

```
In [314]: x = xrange(len(tr_loss + tr_loss2))
    plt.plot(x, tr_loss + tr_loss2, label='train')
    plt.plot(x, tst_loss + tst_loss2, label='test')
    plt.legend(loc='center left', bbox_to_anchor=(1,0.5))
```

Out[314]: <matplotlib.legend.Legend at 0x7f15acde1ed0>



In [315]: @interact(s=widgets.IntSlider(min=0,max=23,step=1,value=0))
 def f(s):
 x = xrange(len((tr_loss + tr_loss2)[s::7]))
 plt.plot(x, (tr_loss + tr_loss2)[s::7], label='train')
 plt.plot(x, (tst_loss + tst_loss2)[s::7], label='test')
 plt.legend(loc='center left', bbox_to_anchor=(1,0.5))



```
In [300]:
          def get NEP prediction(sess, patch tensor, prediction tensor, img,
          centre, H, W, d):
              assert H%2==1
              assert W%2==1
              halfH = (W-1)/2
              halfW = (H-1)/2
               (imgH, imgW, ) = img.shape
               (x, y) = centre
              # Helper function to check for out of bounds
              def inbounds(dx, dy):
                   return (x + dx - halfW >= 0 and
                           x + dx + halfW < imgW and
                           y + dy - halfH >= 0 and
                           y + dy + halfH < imgH)
              # Iterate over each neighbour position
              patches = []
              for dx in range(-d, d+1):
                  for dv in range(-d, d+1):
                      if dx^{**}2 + dy^{**}2 \le d^{**}2 and inbounds(dx, dy):
                           patches.append(img[y+dy-halfH:y+dy+halfW+1,x+dx-halfW:x+d
          x+halfW+1,:])
                           \#print(dx, dy, math.sqrt(dx**2+dy**2))
              assert len(patches) > 0
              patches = np.stack(patches)
              #print patches.shape
              # Run prediction
              predictions = sess.run(prediction tensor, feed dict={
                      patch tensor: patches,
                  })
              # Get average prediction
              #print predictions.shape
              average predictions = np.mean(predictions, axis=0)
              #print average predictions.shape
              return (average predictions, patches)
          example = 3
In [316]:
          (avg pred, patches used) = get NEP prediction(sess,
                                                           patch tensor,
                                                           inference predictions,
                                                           i[test img ids[example]],
                                                           test centres[example],
                                                           27,
                                                           27,
                                                           4)
          print avg pred
```

```
[ 0.90826339  0.03643493  0.04092112  0.01438036]
[1 0 0 0]
```

print test labels[example]

```
In [329]: print sess.run([accuracy, f1, confusion], feed_dict={
                  inference predictions: all nep predictions,
                  label_tensor: test_labels,
              })
          [0.76651114, 0.76562738, array([[1130, 180,
                                                         71,
                                                               12],
                 [ 110, 803, 111,
                                      451,
                         125, 1023,
                 [
                   42,
                                      63],
                    24,
                         77,
                                91,
                                     166]], dtype=int32)]
```

Results for f1-score for various d:

```
0 = .75110334 \ 1 = .75696123 \ 2 = .76257628 \ 3 = .76513195 \ 4 = .76562738
```