

# show\_results

November 3, 2016

## 1 Load model

### 1.1 Model

```
conv1_1 = new_conv_layer( image, [5, 5, 1, 16], "conv1_1" ) conv1_2 = new_conv_layer( conv1_1,
[5, 5, 16, 16], "conv1_2" )
conv2_1 = new_conv_layer(conv1_2, [5, 5, 16, 16], "conv2_1") conv2_2 =
new_conv_layer(conv2_1, [5, 5, 16, 16], "conv2_2")
gap = tf.reduce_mean( conv2_2, [1,2] ) ccn = tf.reshape(gap,[-1,10,nb_CCN]) ccn =
tf.reduce_mean(ccn, 2)
```

### 1.2 train params

L2 on weights (5e-5) L1 on GAP

```
In [1]: import matplotlib.pyplot as plt
import numpy as np
import simple_model
from simple_model import training_generator
import utils

simple_model=reload(simple_model)
lr          = .005
lr_decay    = .9
back_size   = 100
noise       = .1
crop_pos    = (10,10)
n_CCN       = 5
```

Exception AssertionError: AssertionError("Nesting violated for default stack of <ty

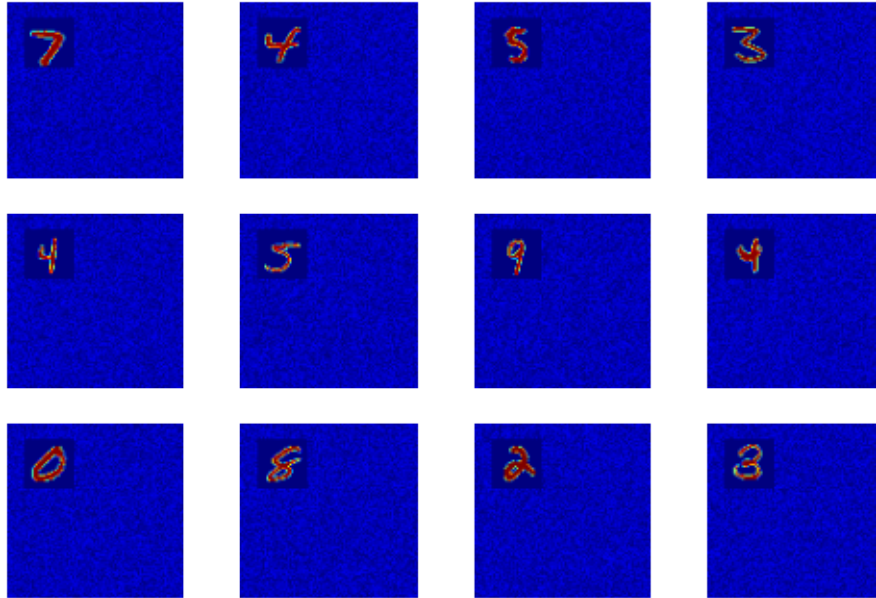
```
In [2]: """Plot training samples"""
batch = utils.get_batch('train', im_size=back_size, noise=noise, crop_pos=c

fig, axs = plt.subplots(3,4)
for ax,img in zip([b for a in axs for b in a],batch[0]):
```

```

        ax.imshow(img.reshape(back_size,back_size), vmin=0, vmax=1)
        ax.set_axis_off()
plt.show()

```



```

In [3]: """Perform training"""
        gen = training_generator(lr=lr, lr_decay=lr_decay, back_size=back_size, noi
        for _ in range(25):
            _, accs = gen.next()
            print "max acc so far : "+str(max(accs)*100)

```

```

***** EPOCH 0 *****
1-Loss on testset is 2.356416
1-Accuracy now is 11.49
2-Loss on testset is 2.301980
2-Accuracy now is 13.02
lr now is 0.00450
max acc so far : 11.49

```

```

***** EPOCH 1 *****
1-Loss on testset is 1.360155
1-Accuracy now is 75.79
2-Loss on testset is 1.361032
2-Accuracy now is 75.69
lr now is 0.00405

```

max acc so far : 75.79

\*\*\*\*\* EPOCH 2 \*\*\*\*\*

1-Loss on testset is 0.912301

1-Accuracy now is 88.79

2-Loss on testset is 0.916553

2-Accuracy now is 88.50

lr now is 0.00365

max acc so far : 88.79

\*\*\*\*\* EPOCH 3 \*\*\*\*\*

1-Loss on testset is 0.768343

1-Accuracy now is 90.11

2-Loss on testset is 0.766207

2-Accuracy now is 90.12

lr now is 0.00328

max acc so far : 90.11

\*\*\*\*\* EPOCH 4 \*\*\*\*\*

1-Loss on testset is 0.680467

1-Accuracy now is 90.77

2-Loss on testset is 0.673926

2-Accuracy now is 90.96

lr now is 0.00295

max acc so far : 90.77

\*\*\*\*\* EPOCH 5 \*\*\*\*\*

1-Loss on testset is 0.618553

1-Accuracy now is 91.27

2-Loss on testset is 0.616227

2-Accuracy now is 91.36

lr now is 0.00266

max acc so far : 91.27

\*\*\*\*\* EPOCH 6 \*\*\*\*\*

1-Loss on testset is 0.504258

1-Accuracy now is 95.03

2-Loss on testset is 0.500635

2-Accuracy now is 95.07

lr now is 0.00239

max acc so far : 95.03

\*\*\*\*\* EPOCH 7 \*\*\*\*\*

1-Loss on testset is 0.472787

1-Accuracy now is 95.69

2-Loss on testset is 0.469958

2-Accuracy now is 95.63

lr now is 0.00215

max acc so far : 95.69

\*\*\*\*\* EPOCH 8 \*\*\*\*\*

1-Loss on testset is 0.457466

1-Accuracy now is 95.85

2-Loss on testset is 0.448033

2-Accuracy now is 96.15

lr now is 0.00194

max acc so far : 95.85

\*\*\*\*\* EPOCH 9 \*\*\*\*\*

1-Loss on testset is 0.437950

1-Accuracy now is 95.90

2-Loss on testset is 0.442658

2-Accuracy now is 95.66

lr now is 0.00174

max acc so far : 95.9

\*\*\*\*\* EPOCH 10 \*\*\*\*\*

1-Loss on testset is 0.409138

1-Accuracy now is 96.80

2-Loss on testset is 0.402230

2-Accuracy now is 96.91

lr now is 0.00157

max acc so far : 96.8

\*\*\*\*\* EPOCH 11 \*\*\*\*\*

1-Loss on testset is 0.402402

1-Accuracy now is 96.80

2-Loss on testset is 0.399086

2-Accuracy now is 96.84

lr now is 0.00141

max acc so far : 96.8

\*\*\*\*\* EPOCH 12 \*\*\*\*\*

1-Loss on testset is 0.399370

1-Accuracy now is 97.21

2-Loss on testset is 0.396558

2-Accuracy now is 97.10

lr now is 0.00127

max acc so far : 97.21

\*\*\*\*\* EPOCH 13 \*\*\*\*\*

1-Loss on testset is 0.413590

1-Accuracy now is 96.19

2-Loss on testset is 0.400492

2-Accuracy now is 96.57

lr now is 0.00114

max acc so far : 97.21

\*\*\*\*\* EPOCH 14 \*\*\*\*\*

1-Loss on testset is 0.386078

1-Accuracy now is 96.76

2-Loss on testset is 0.387164

2-Accuracy now is 96.75

lr now is 0.00103

max acc so far : 97.21

\*\*\*\*\* EPOCH 15 \*\*\*\*\*

1-Loss on testset is 0.384180

1-Accuracy now is 97.20

2-Loss on testset is 0.380877

2-Accuracy now is 97.32

lr now is 0.00093

max acc so far : 97.21

\*\*\*\*\* EPOCH 16 \*\*\*\*\*

1-Loss on testset is 0.363440

1-Accuracy now is 97.24

2-Loss on testset is 0.355429

2-Accuracy now is 97.40

lr now is 0.00083

max acc so far : 97.24

\*\*\*\*\* EPOCH 17 \*\*\*\*\*

1-Loss on testset is 0.373458

1-Accuracy now is 97.15

2-Loss on testset is 0.362167

2-Accuracy now is 97.27

lr now is 0.00075

max acc so far : 97.24

\*\*\*\*\* EPOCH 18 \*\*\*\*\*

1-Loss on testset is 0.359458

1-Accuracy now is 97.58

2-Loss on testset is 0.352296

2-Accuracy now is 97.70

lr now is 0.00068

max acc so far : 97.58

\*\*\*\*\* EPOCH 19 \*\*\*\*\*

1-Loss on testset is 0.355177

1-Accuracy now is 97.52

2-Loss on testset is 0.348388

2-Accuracy now is 97.49

lr now is 0.00061

max acc so far : 97.58

\*\*\*\*\* EPOCH 20 \*\*\*\*\*

1-Loss on testset is 0.347565

1-Accuracy now is 97.45

2-Loss on testset is 0.339475

2-Accuracy now is 97.62

lr now is 0.00055

max acc so far : 97.58

\*\*\*\*\* EPOCH 21 \*\*\*\*\*

1-Loss on testset is 0.351418

1-Accuracy now is 97.66

2-Loss on testset is 0.346345

2-Accuracy now is 97.66

lr now is 0.00049

max acc so far : 97.66

\*\*\*\*\* EPOCH 22 \*\*\*\*\*

1-Loss on testset is 0.349502

1-Accuracy now is 97.56

2-Loss on testset is 0.346330

2-Accuracy now is 97.52

lr now is 0.00044

max acc so far : 97.66

\*\*\*\*\* EPOCH 23 \*\*\*\*\*

1-Loss on testset is 0.350790

1-Accuracy now is 97.73

2-Loss on testset is 0.343002

2-Accuracy now is 97.77

lr now is 0.00040

max acc so far : 97.73

\*\*\*\*\* EPOCH 24 \*\*\*\*\*

1-Loss on testset is 0.345823

1-Accuracy now is 97.66

2-Loss on testset is 0.336625

2-Accuracy now is 97.68

lr now is 0.00036

max acc so far : 97.73

In [4]: back\_size = 100

```
for _ in range(10):
```

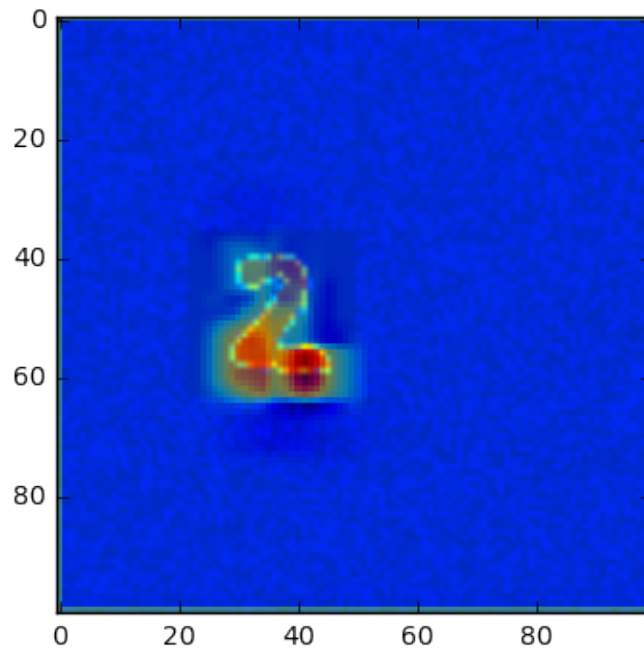
```
    print '-----'*5
```

```
    imgs, lbls = utils.get_batch('test', 1, back_size, .1).next()
```

```
simple_model.show_activation(imgs[0])
```

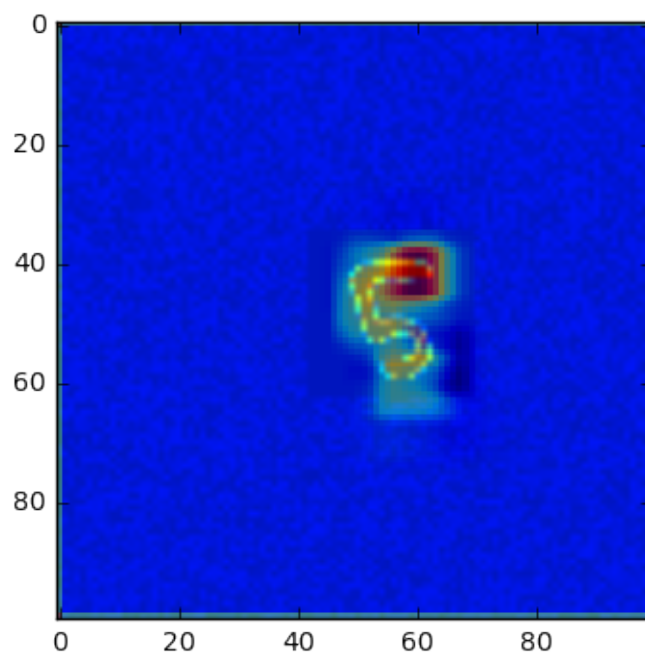
---

```
prediction is : 2 with 7.963
```



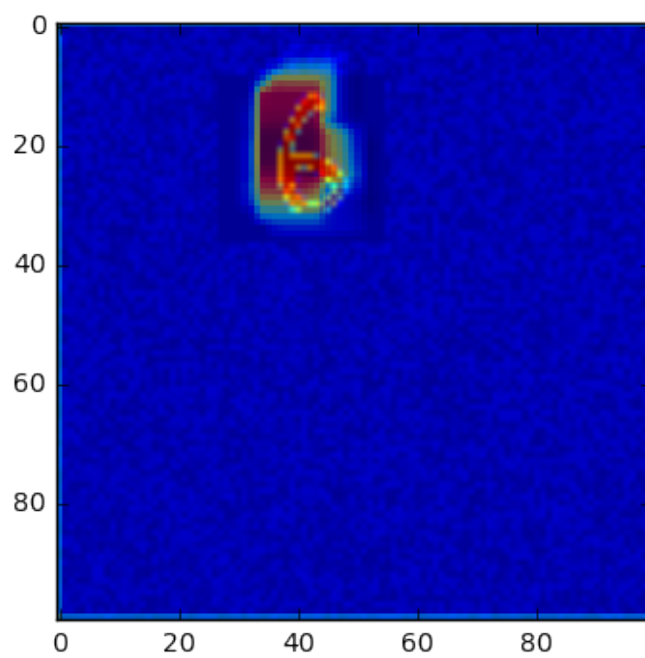
---

```
prediction is : 5 with 12.150
```



---

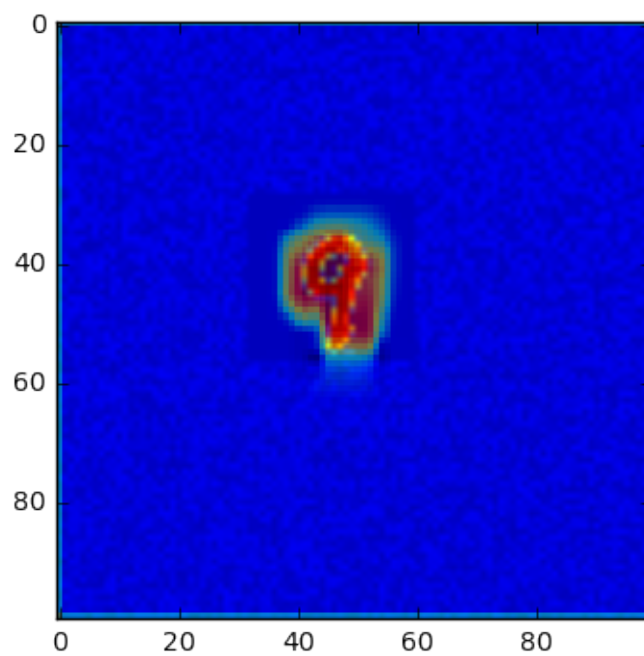
prediction is : 6 with 7.511





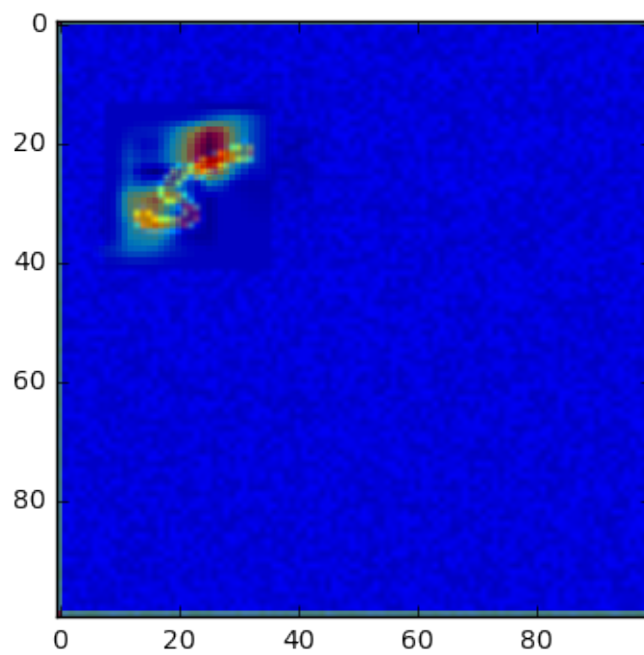
---

prediction is : 9 with 8.456



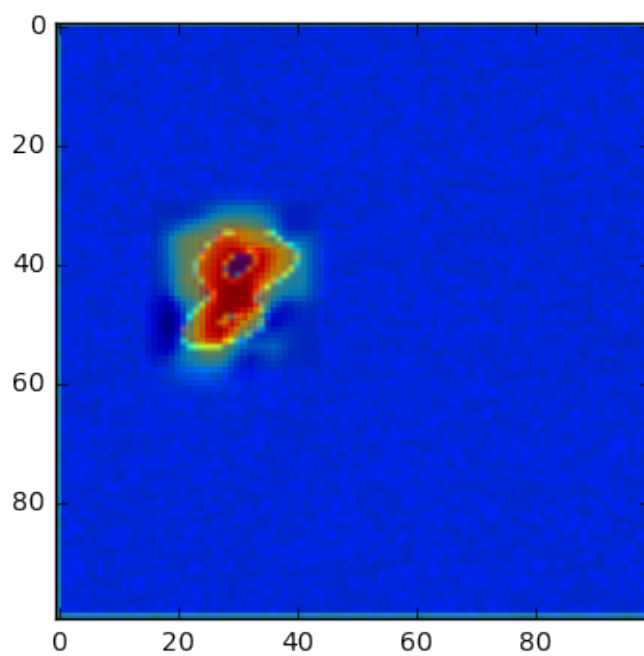
---

prediction is : 5 with 9.257



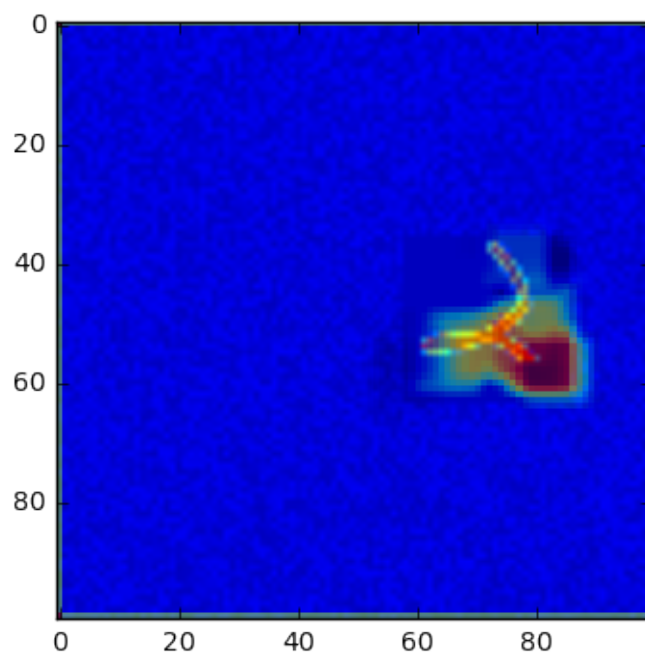
---

prediction is : 8 with 10.372

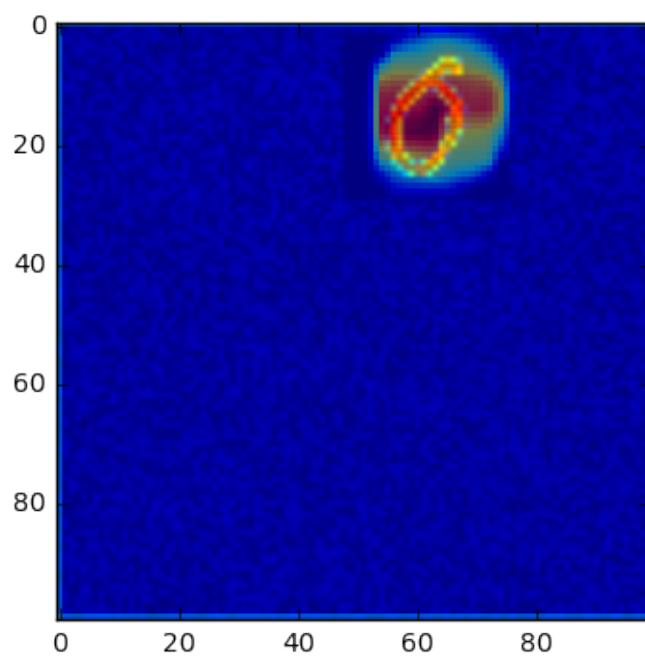


---

prediction is : 2 with 6.469

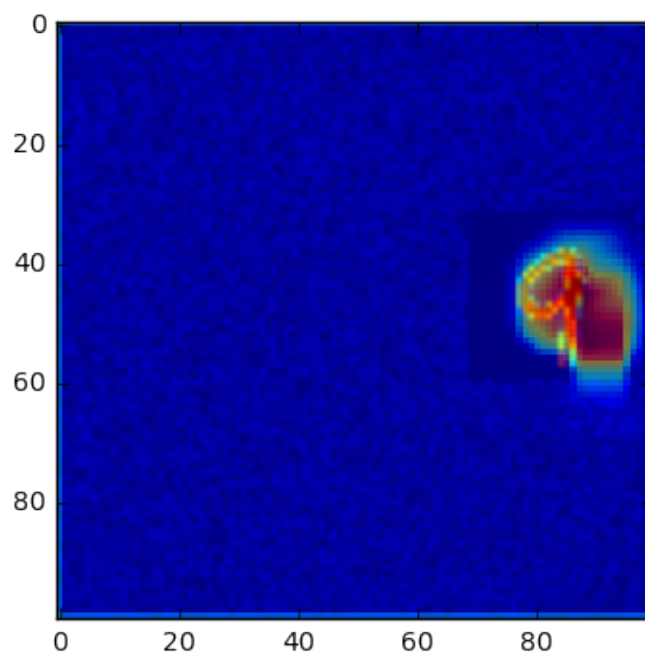


-----  
prediction is : 0 with 5.794



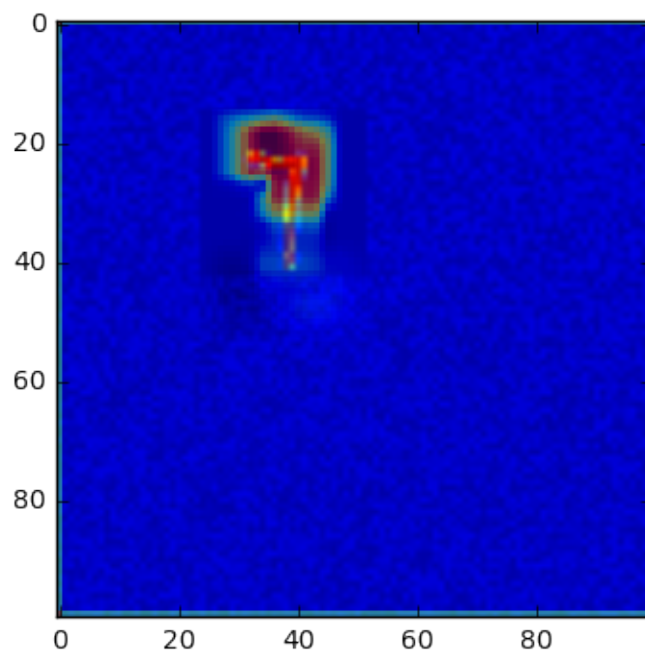
---

prediction is : 9 with 8.243



---

prediction is : 7 with 6.853



```

In [5]: accuracies = []
        for _ in range(100):
            digit, lbl = utils.get_batch('test', 1, .1).next()
            digit      = digit[0].reshape(28,28)
            img        = np.random.random((100,100))*0.1
            height     = img.shape[0]
            width      = img.shape[1]
            box_size   = 28
            step_size  = 10
            n_x_boxes  = (width - box_size) / step_size + 1
            n_y_boxes  = (height - box_size) / step_size + 1
            imgs       = np.tile(img, (n_x_boxes*n_y_boxes,1,1))

            for xx in range(0, n_x_boxes ):
                for yy in range(0, n_y_boxes ):
                    idx = xx*n_x_boxes+yy
                    x    = xx*step_size
                    y    = yy*step_size
                    imgs[idx, x:x+box_size, y:y+box_size ] = digit

            imgs = imgs.reshape((-1,100,100,1))
            preds = simple_model.sess.run(simple_model.tf_out, feed_dict={simple_model.x: imgs})
            pred  = np.argmax(preds[-1])

            accuracy = sum(preds.argmax(axis=1) == lbl) / float(len(preds))
            accuracies.append(accuracy)

        print sum(accuracies)/len(accuracies)

```

1.0

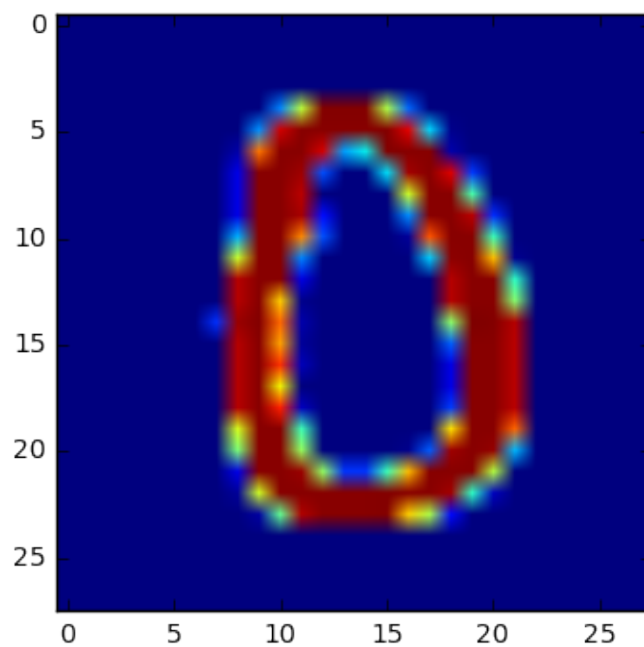
```

In [6]: digit, lbl = utils.get_batch('test', 1, .1).next()
        digit      = digit[0].reshape(28,28)
        img        = np.random.random((100,100))*0.1
        height     = img.shape[0]
        width      = img.shape[1]
        box_size   = 28
        step_size  = 10
        n_x_boxes  = (width - box_size) / step_size + 1
        n_y_boxes  = (height - box_size) / step_size + 1
        imgs       = np.tile(img, (n_x_boxes*n_y_boxes,1,1))

        for xx in range(0, n_x_boxes ):

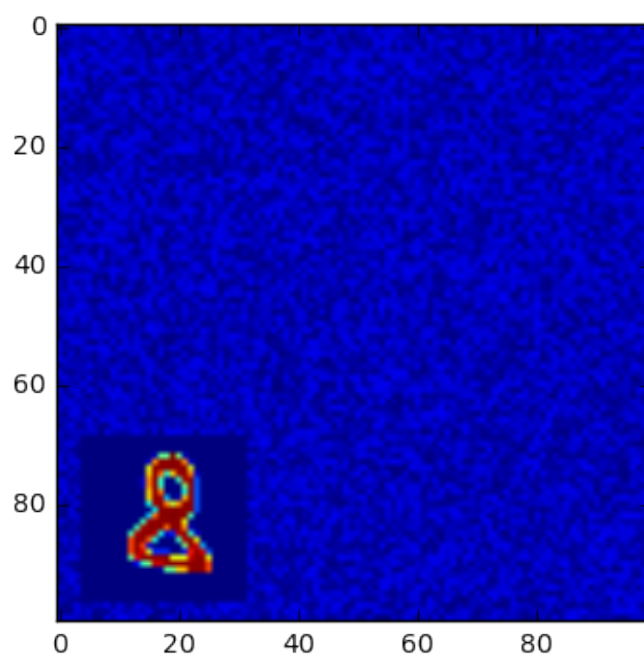
```

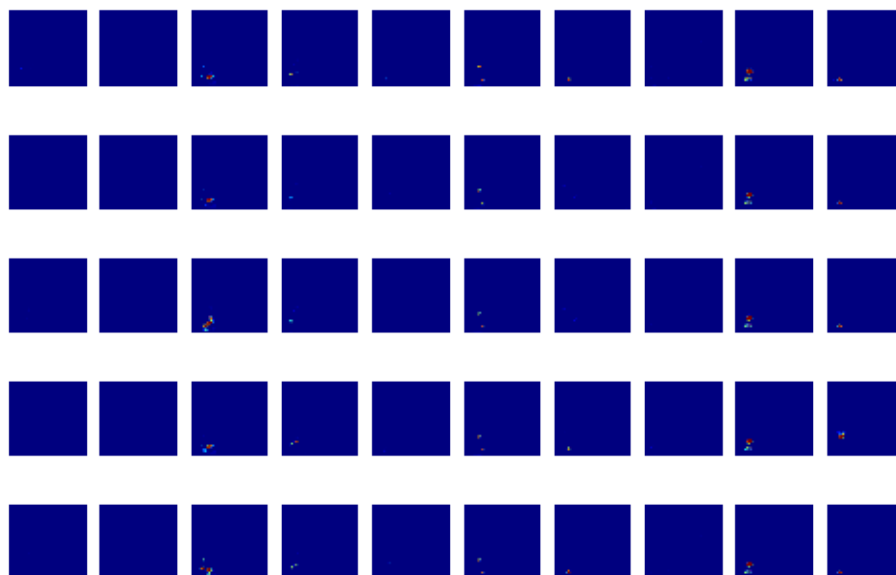




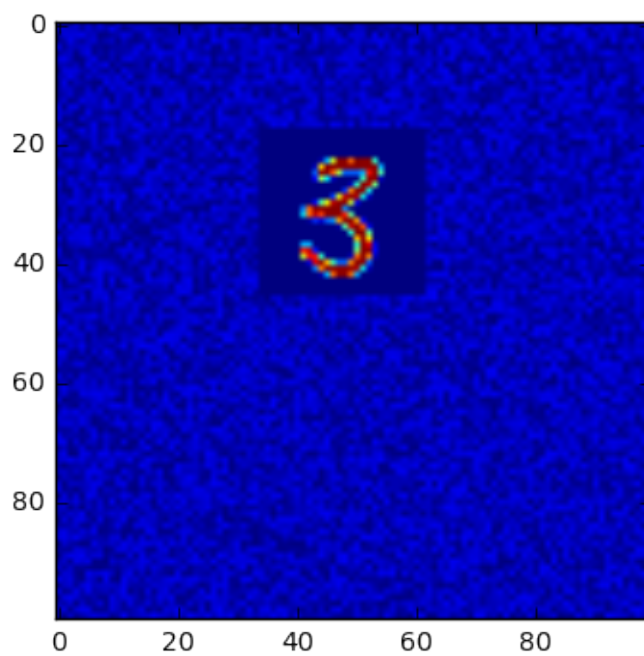
```
In [7]: for _ in range(10):
         simple_model.plot_classes_maps(100, n_CCN)
```

8

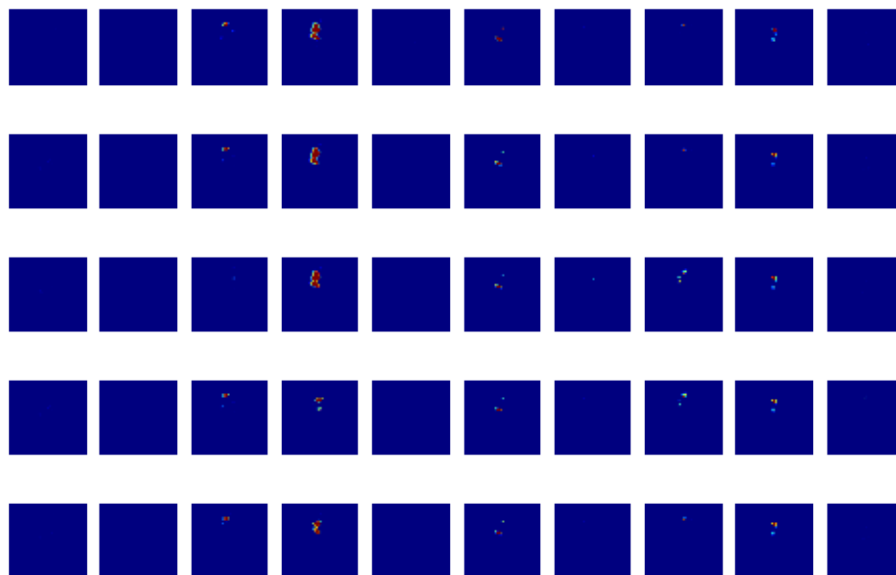




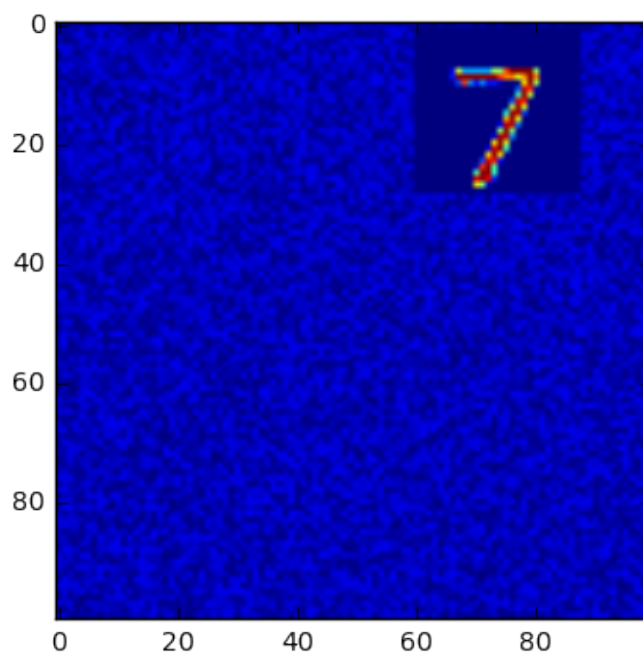
3





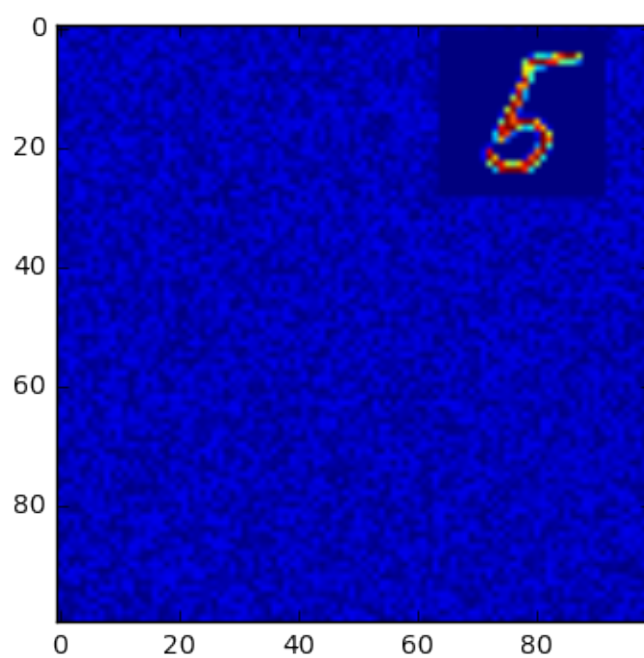


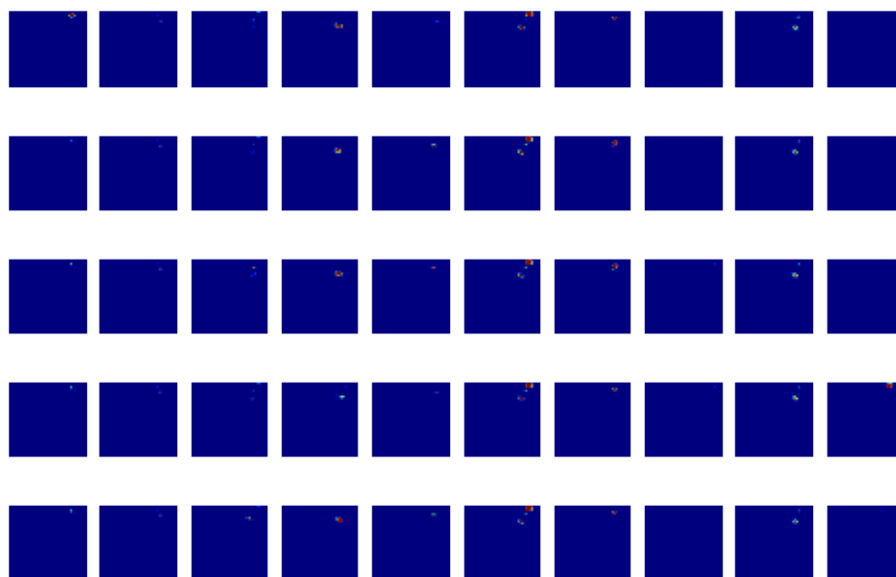
7



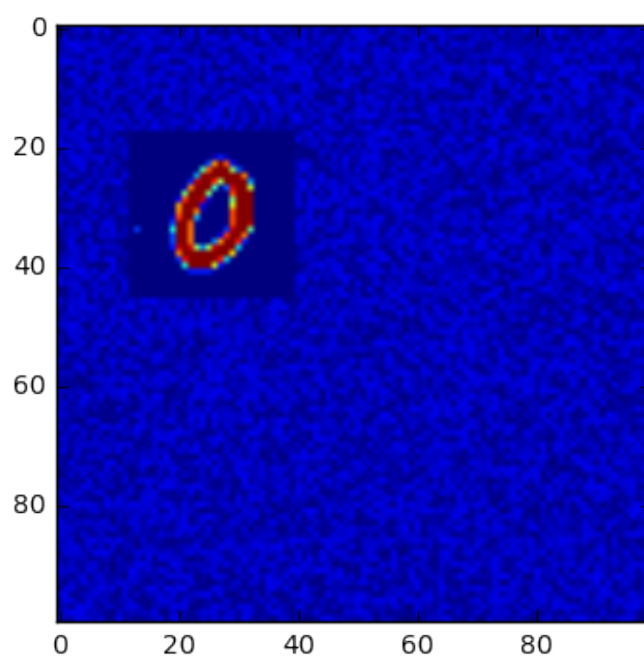


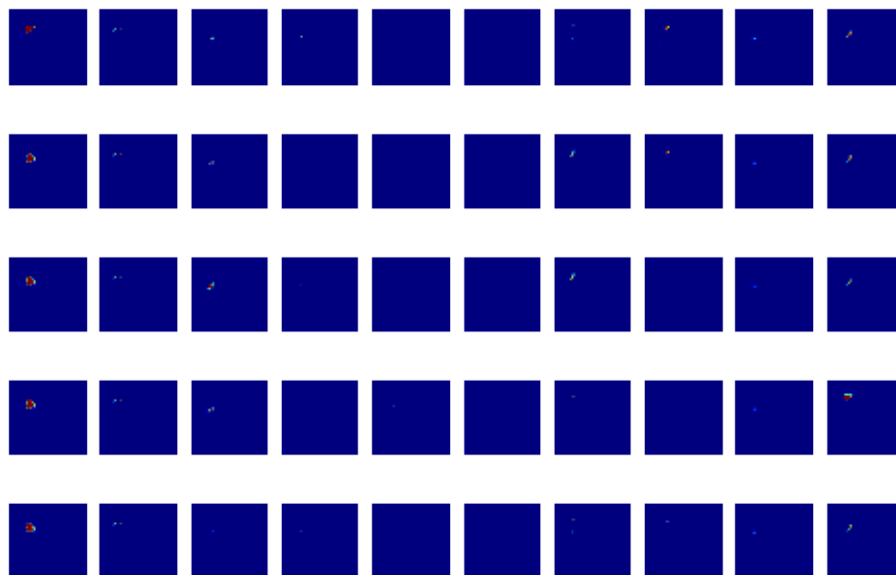
5



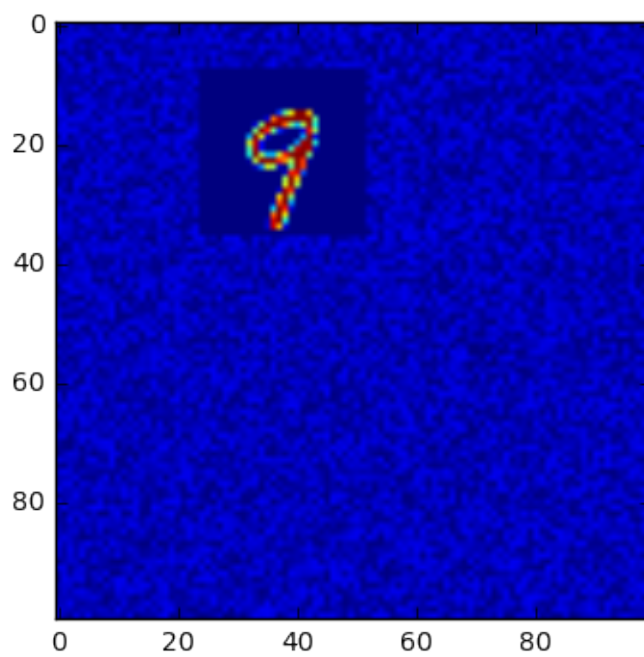


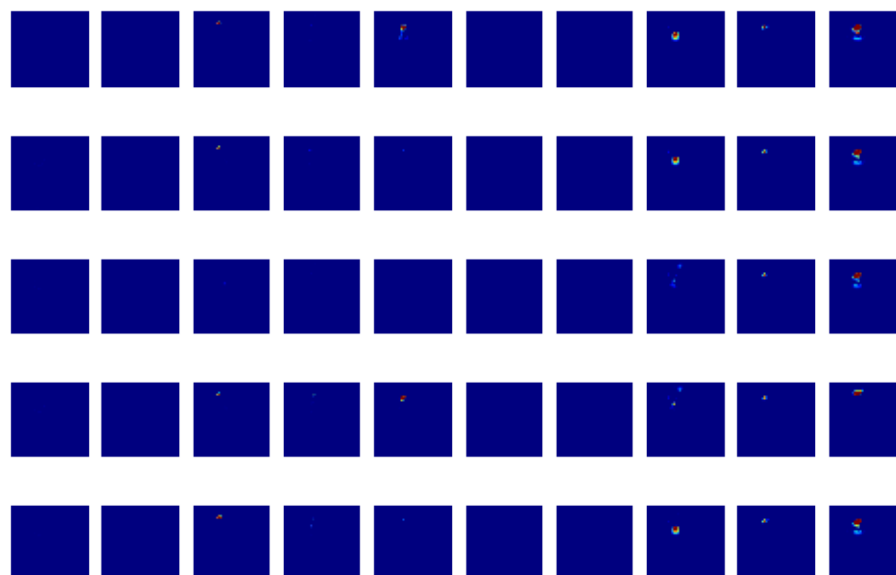
0



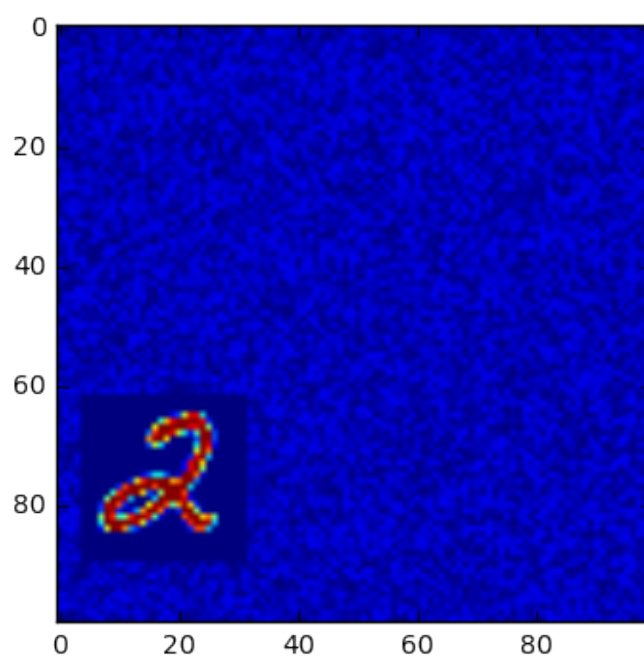


9



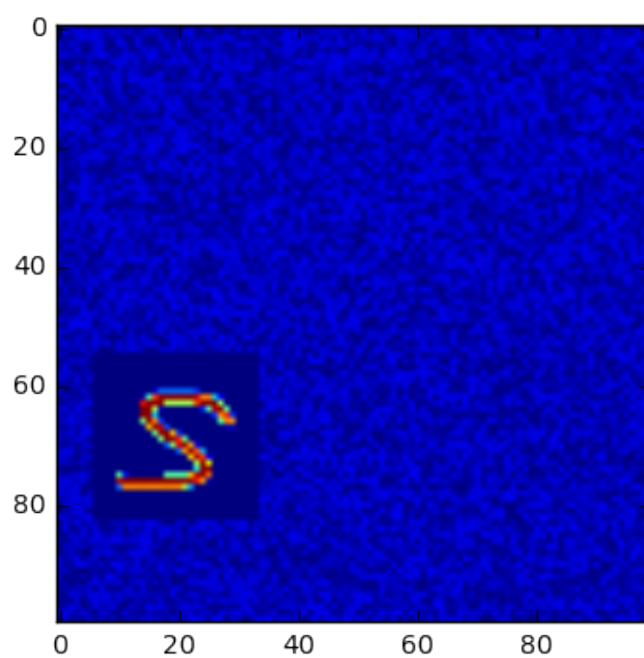


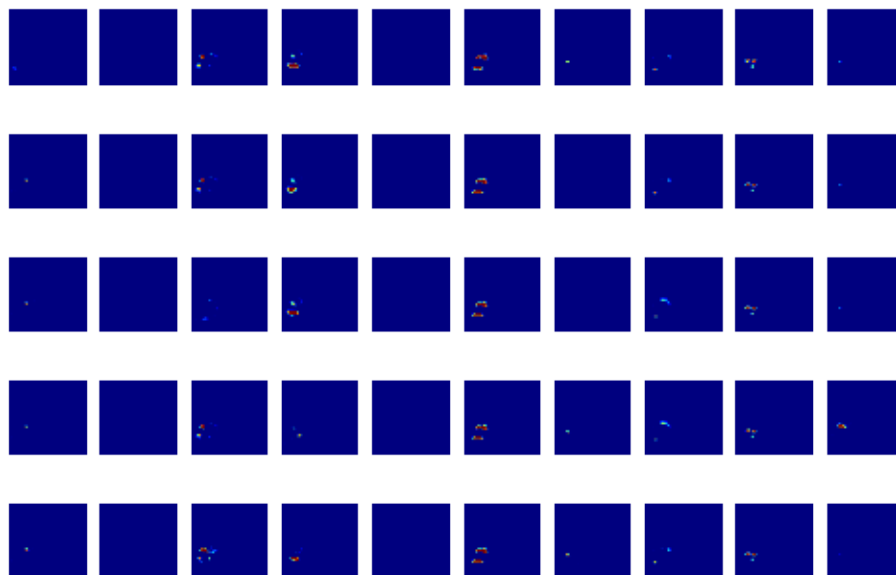
2



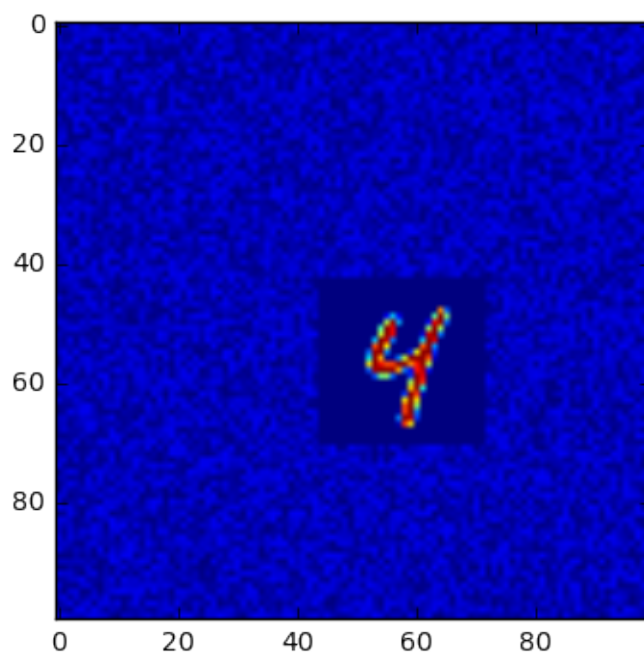


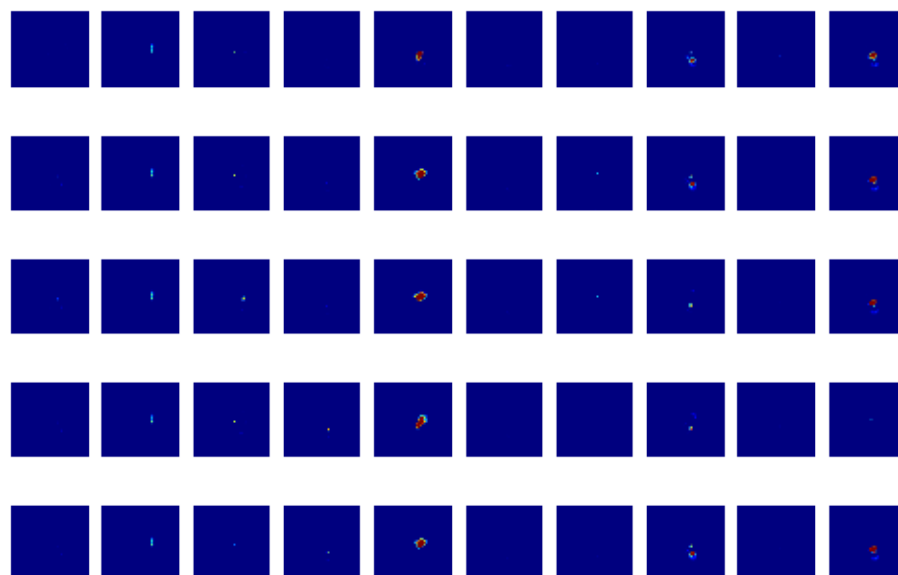
5



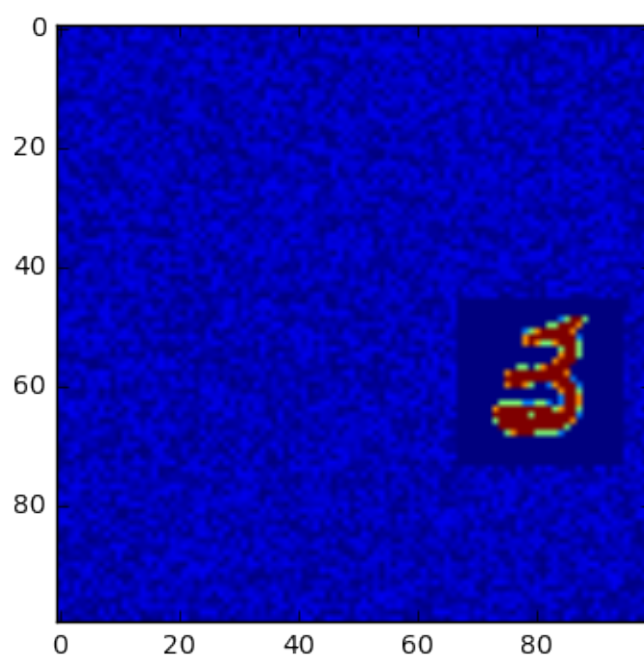


4

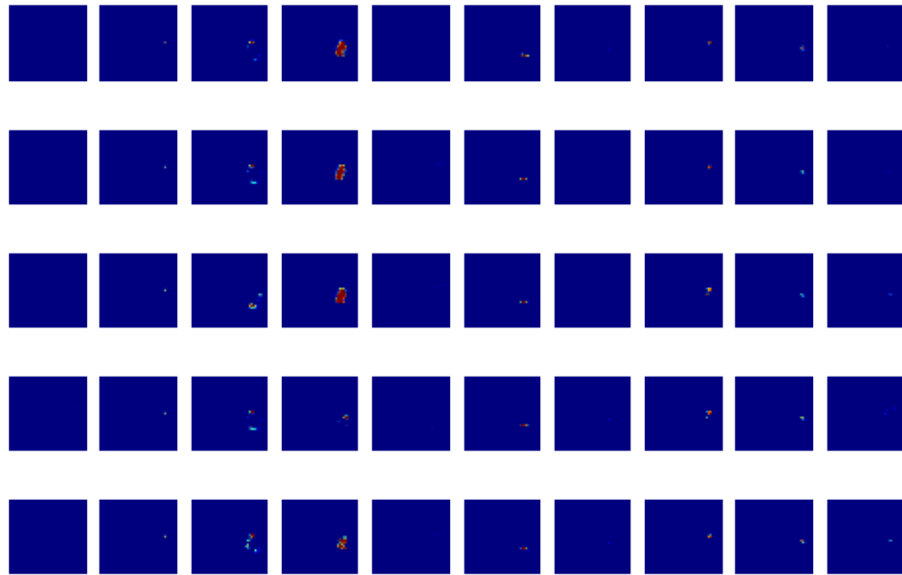




3







In [ ]: