

# hw2\_p1

February 16, 2018

```
In [1]: import numpy as np
import tensorflow as tf
%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt
```

```
In [2]: from tensorflow.examples.tutorials.mnist import input_data
mnist = input_data.read_data_sets("MNIST_data/", one_hot=False)
print(mnist.train.images.shape, mnist.train.labels.shape)
```

```
Extracting MNIST_data/train-images-idx3-ubyte.gz
Extracting MNIST_data/train-labels-idx1-ubyte.gz
Extracting MNIST_data/t10k-images-idx3-ubyte.gz
Extracting MNIST_data/t10k-labels-idx1-ubyte.gz
(55000, 784) (55000,)
```

```
In [3]: def showimage(image, label):
    plt.gray()
    plt.imshow(image.reshape(28, 28))
    plt.show()
    print(label)

def extract_target_data(X, Y, target, num):
    p = (Y == target)
    x_target = X[p,:]
    y_target = Y[p]
    y_target = np.expand_dims(y_target, axis=1)
    return x_target[:num,:], y_target[:num,:]

# computer euclidean distance matrix
def euclidean_distance_matrix(x):
    r = tf.reduce_sum(x*x, 1)
    r = tf.reshape(r, [-1, 1])
    distance_mat = r - 2*tf.matmul(x, tf.transpose(x)) + tf.transpose(r)
    #return tf.sqrt(distance_mat)
    return distance_mat
```

```

In [4]: train_images, train_labels = extract_target_data(mnist.train.images, mnist.train.labels,

# get 1000 data from each category
for i in range(1, 10):
    cur_train_images, cur_train_labels = extract_target_data(mnist.train.images, mnist.t
    train_images = np.concatenate((train_images, cur_train_images), axis=0)
    train_labels = np.concatenate((train_labels, cur_train_labels), axis=0)

train_images /= 255.
#print(train_images.shape, train_labels.shape)

In [5]: distance_mat = euclidean_distance_matrix(train_images)
with tf.Session() as sess:
    M = sess.run(distance_mat)

In [6]: with tf.device('/device:GPU:0'):
    x = tf.placeholder(tf.float32, shape=[None, 784], name='x')
    w = tf.get_variable('w', shape=[784, 2], initializer=tf.contrib.layers.xavier_initia
    b = tf.get_variable('b', shape=[2], initializer=tf.zeros_initializer())

    z = tf.matmul(x, w) + b
    M_ = euclidean_distance_matrix(z)
    # MSE cost function
    cost = tf.reduce_mean(tf.square(M_ - M))
    optimizer = tf.train.AdamOptimizer().minimize(cost)

In [7]: epoch = 1000
config = tf.ConfigProto()
config.gpu_options.allow_growth = True
with tf.Session(config=config) as sess:
    sess.run(tf.global_variables_initializer())
    for i in range(epoch):
        _, epoch_cost, transform_mat = sess.run([optimizer, cost, w], feed_dict = {x:tra

In [8]: emb_images = np.dot(train_images, transform_mat)
print(emb_images.shape)

(10000, 2)

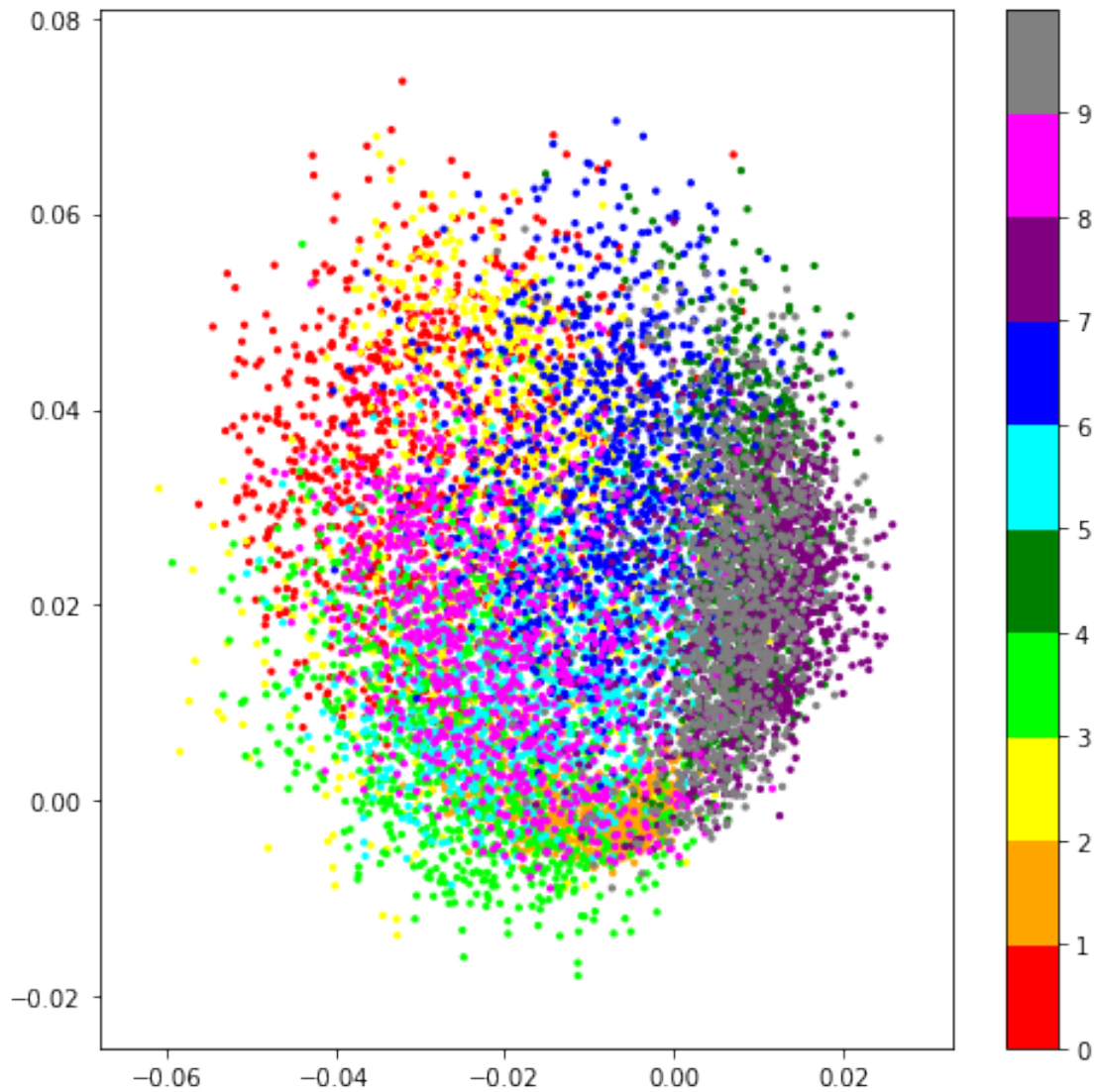
In [9]: x = emb_images[:,0]
y = emb_images[:,1]
label = train_labels
colors = ['red', 'orange', 'yellow', 'lime', 'green', 'cyan', 'blue', 'purple', 'magenta', 'grey'

fig = plt.figure(figsize=(8,8))
plt.scatter(x, y, c=label, cmap=matplotlib.colors.ListedColormap(colors),s=5)

cb = plt.colorbar()

```

```
loc = np.arange(0,max(label),max(label)/float(len(colors)))
cb.set_ticks(loc)
cb.set_ticklabels(list(range(10)))
```



We can see from the image above, the datapoints of same category are clustering together. This infers the embedding, even though with only 2 dimensions, preserve essential distance information from 786 dimensional matrix and are able to dissimilate from other categories.

# hw2\_p2

February 16, 2018

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf

%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt

import pymesh
from pyntcloud import PyntCloud

In [2]: # compute triangle mesh surface area
def triangle_area(x):
    a = x[:,0,:] - x[:,1,:]
    b = x[:,0,:] - x[:,2,:]
    cross = np.cross(a, b)
    area = 0.5 * np.linalg.norm(np.cross(a, b), axis=1)
    return area

# compute euclidean distance matrix
def euclidean_distance_matrix(x):
    r = np.sum(x*x, 1)
    r = r.reshape(-1, 1)
    distance_mat = r - 2*np.dot(x, x.T) + r.T
    #return np.sqrt(distance_mat)
    return distance_mat

# update distance matrix and select the farthest point from set S after a new point is s
def update_farthest_distance(far_mat, dist_mat, s):
    for i in range(far_mat.shape[0]):
        far_mat[i] = dist_mat[i,s] if far_mat[i] > dist_mat[i,s] else far_mat[i]
    return far_mat, np.argmax(far_mat)

# initialize matrix to keep track of distance from set s
def init_farthest_distance(far_mat, dist_mat, s):
    for i in range(far_mat.shape[0]):
        far_mat[i] = dist_mat[i,s]
    return far_mat
```

```

In [3]: # get sample from farthest point on every iteration
def farthest_point_sampling(obj_file, num_samples=1000):
    mesh = pymesh.load_mesh(obj_file)
    faces = mesh.vertices[mesh.faces]
    area = triangle_area(faces)
    total_area = np.sum(area)

    set_P = []
    for i in range(faces.shape[0]):
        num_gen = area[i] / total_area * 10000
        for j in range(int(num_gen)+1):
            r1, r2 = np.random.rand(2)
            d = (1-np.sqrt(r1)) * faces[i,0] + np.sqrt(r1)*(1-r2) * faces[i,1] + np.sqrt(r2) * faces[i,2]
            set_P.append(d)

    set_P = np.array(set_P)
    num_P = set_P.shape[0]

    distance_mat = euclidean_distance_matrix(set_P)

    set_S = []
    s = np.random.randint(num_P)
    far_mat = init_farthest_distance(np.zeros((num_P)), distance_mat, s)

    for i in range(num_samples):
        set_S.append(set_P[s])
        far_mat, s = update_farthest_distance(far_mat, distance_mat, s)

    return np.array(set_S)

In [4]: teapot_pts = farthest_point_sampling('teapot.obj')

In [5]: points = pd.DataFrame(teapot_pts, columns=['x', 'y', 'z'])
        cloud = PyntCloud(points)
        cloud.plot(line_color='')

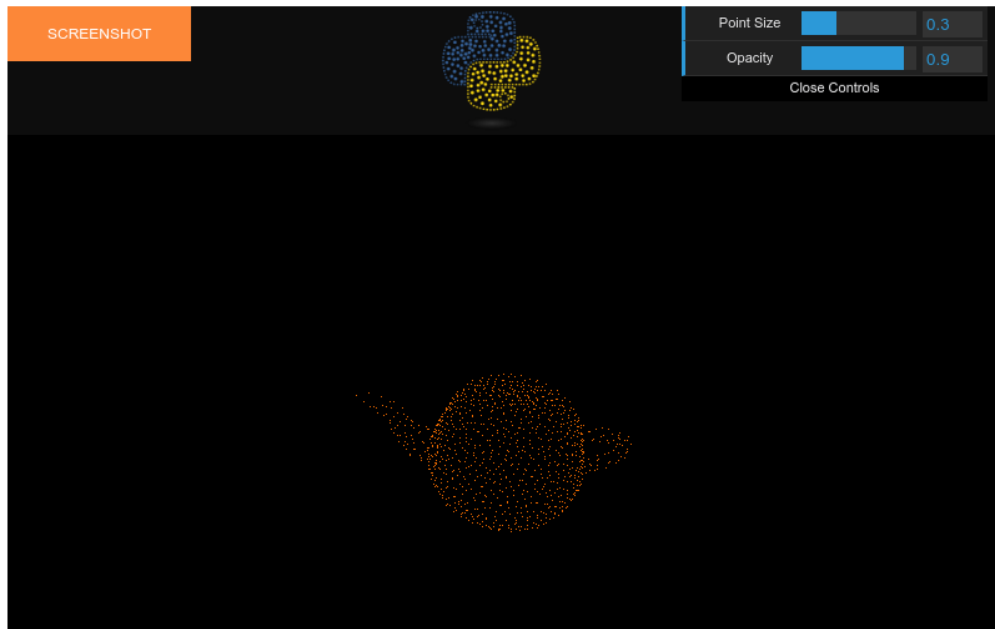
Out[5]: <IPython.lib.display.IFrame at 0x7fc5efb855f8>

In [6]: violin_pts = farthest_point_sampling('violin_case.obj')

In [7]: points = pd.DataFrame(violin_pts, columns=['x', 'y', 'z'])
        cloud = PyntCloud(points)
        cloud.plot(line_color='')

Out[7]: <IPython.lib.display.IFrame at 0x7fc5ef922630>

```



teapot



violin case

# hw2\_p3

February 17, 2018

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf
import time
from scipy.optimize import linear_sum_assignment

%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt

import pymesh
from pyntcloud import PyntCloud

from tf_emddistance import emd_distance

In [2]: # This section of code is copied from problem 2
def triangle_area(x):
    a = x[:,0,:] - x[:,1,:]
    b = x[:,0,:] - x[:,2,:]
    cross = np.cross(a, b)
    area = 0.5 * np.linalg.norm(np.cross(a, b), axis=1)
    return area

def euclidean_distance_matrix(x):
    r = np.sum(x*x, 1)
    r = r.reshape(-1, 1)
    distance_mat = r - 2*np.dot(x, x.T) + r.T
    #return np.sqrt(distance_mat)
    return distance_mat

def update_farthest_distance(far_mat, dist_mat, s):
    for i in range(far_mat.shape[0]):
        far_mat[i] = dist_mat[i,s] if far_mat[i] > dist_mat[i,s] else far_mat[i]
    return far_mat, np.argmax(far_mat)

def init_farthest_distance(far_mat, dist_mat, s):
    for i in range(far_mat.shape[0]):
```

```

        far_mat[i] = dist_mat[i,s]
    return far_mat

def farthest_point_sampling(obj_file, num_samples=1000):
    mesh = pymesh.load_mesh(obj_file)
    faces = mesh.vertices[mesh.faces]
    area = triangle_area(faces)
    total_area = np.sum(area)

    set_P = []
    for i in range(faces.shape[0]):
        num_gen = area[i] / total_area * 10000
        for j in range(int(num_gen)+1):
            r1, r2 = np.random.rand(2)
            d = (1-np.sqrt(r1)) * faces[i,0] + np.sqrt(r1)*(1-r2) * faces[i,1] + np.sqrt(r2) * faces[i,2]
            set_P.append(d)

    set_P = np.array(set_P)
    num_P = set_P.shape[0]

    distance_mat = euclidean_distance_matrix(set_P)

    set_S = []
    s = np.random.randint(num_P)
    far_mat = init_farthest_distance(np.zeros((num_P)), distance_mat, s)

    for i in range(num_samples):
        set_S.append(set_P[s])
        far_mat, s = update_farthest_distance(far_mat, distance_mat, s)

    return np.array(set_S, dtype=np.float32)

def pointcloud_distance_matrix(x1, x2):
    a = tf.expand_dims(x1, axis=1)
    b = tf.expand_dims(x2, axis=0)
    distance_mat = tf.norm(a - b, axis=-1)
    return distance_mat

In [3]: # Sample cloud points -- same procedure as problem 2
teapot_pts = farthest_point_sampling('../teapot.obj', 200)
violin_pts = farthest_point_sampling('../violin_case.obj', 200)

In [4]: # compute distance matrix between 2 cloud points
with tf.Session() as sess:
    distance_mat = sess.run(pointcloud_distance_matrix(teapot_pts, violin_pts))

In [5]: class hungary():

```



```

def __init__(self, weight):
    self.n = weight.shape[0]

    self.w = np.copy(weight)
    # cost matrix
    self.c = np.copy(weight)
    self.m = np.zeros((self.n, self.n), dtype=int)

    # record row and col covers
    self.RowCover = np.zeros((self.n), dtype=bool)
    self.ColCover = np.zeros((self.n), dtype=bool)
    # record augment paths
    self.path = np.zeros((2*self.n, 2), dtype=int)

# main program, run the algo through steps
def run_hungry(self):
    done = False
    step = 1
    while not done:
        if step == 1:
            step = self.step1()
        elif step == 2:
            step = self.step2()
        elif step == 3:
            step = self.step3()
        elif step == 4:
            step = self.step4()
        elif step == 5:
            step = self.step5()
        elif step == 6:
            step = self.step6()
        elif step == 7:
            done = True

# Each row subtract smallest elements
def step1(self):
    self.c -= np.min(self.c, axis=1, keepdims=True)
    return 2

# star zeros
def step2(self):
    for u in range(self.n):
        for v in range(self.n):
            if self.c[u,v] == 0 and not self.RowCover[u] and not self.ColCover[v]:
                self.m[u, v] = 1
                self.RowCover[u] = True
                self.ColCover[v] = True
                break

```

```

        self.clear_covers()
        return 3

# cover cols with starred zeros. check if done
def step3(self):
    for u in range(self.n):
        for v in range(self.n):
            if self.m[u, v] == 1:
                self.ColCover[v] = True

    colcnt = np.sum(self.ColCover)

    if colcnt >= self.n:
        return 7
    else:
        return 4

# find noncovered zero and prime it (starred as 2)
def step4(self):
    while True:
        row, col = self.find_a_zero()
        if row == -1:
            return 6
        else:
            self.m[row, col] = 2
            if self.star_in_row(row):
                col = self.find_star_in_row(row)
                self.RowCover[row] = True
                self.ColCover[col] = False
            else:
                self.path_row_0 = row
                self.path_col_0 = col
                return 5

# use augment algo to increase matches
def step5(self):
    done = False
    self.path_count = 1
    self.path[self.path_count-1, 0] = self.path_row_0
    self.path[self.path_count-1, 1] = self.path_col_0

    while not done:
        row = self.find_star_in_col(self.path[self.path_count-1, 1])
        if row > -1:
            self.path_count += 1
            self.path[self.path_count-1, 0] = row
            self.path[self.path_count-1, 1] = self.path[self.path_count-2, 1]
        else:

```

```

        done = True
    if not done:
        col = self.find_prime_in_row(self.path[self.path_count-1, 0])
        self.path_count += 1
        self.path[self.path_count-1, 0] = self.path[self.path_count-2, 0]
        self.path[self.path_count-1, 1] = col

    self.augment_path()
    self.clear_covers()
    self.erase_prime()
    return 3

# add minval val to double covered elements and subtract it to noncovered elements
def step6(self):
    minval = self.find_smallest()
    for u in range(self.n):
        for v in range(self.n):
            if self.RowCover[u]:
                self.c[u,v] += minval
            if not self.ColCover[v]:
                self.c[u,v] -= minval
    return 4

# find first uncovered zero
def find_a_zero(self):
    for u in range(self.n):
        for v in range(self.n):
            if self.c[u,v] == 0 and not self.RowCover[u] and not self.ColCover[v]:
                return u, v
    return -1, -1

def star_in_row(self, row):
    for v in range(self.n):
        if self.m[row, v] == 1:
            return True
    return False

def find_star_in_row(self, row):
    for v in range(self.n):
        if self.m[row, v] == 1:
            return v
    return -1

def find_star_in_col(self, col):
    for u in range(self.n):
        if self.m[u, col] == 1:
            return u
    return -1

```

```

def find_prime_in_row(self, row):
    for v in range(self.n):
        if self.m[row, v] == 2:
            return v
    return -1

def augment_path(self):
    for p in range(self.path_count):
        if self.m[self.path[p,0], self.path[p,1]] == 1:
            self.m[self.path[p,0], self.path[p,1]] = 0
        else:
            self.m[self.path[p,0], self.path[p,1]] = 1

def clear_covers(self):
    self.RowCover = np.zeros((self.n), dtype=bool)
    self.ColCover = np.zeros((self.n), dtype=bool)

def erase_prime(self):
    for u in range(self.n):
        for v in range(self.n):
            if self.m[u,v] == 2:
                self.m[u,v] = 0

def find_smallest(self):
    minval = np.max(self.c)
    for u in range(self.n):
        for v in range(self.n):
            if self.c[u,v] < minval and not self.RowCover[u] and not self.ColCover[v]:
                minval = self.c[u,v]
    return minval

```

```

In [7]: H = hungary(distance_mat)
        H.run_hungary()

```

```

In [8]: # EMD computed by my hungarian algorithm
        np.sum(H.w * H.m)

```

```

Out[8]: 10703.283664703369

```

```

In [9]: # EMD computed by scipy linear_sum_assignment
        row_ind, col_ind = linear_sum_assignment(distance_mat)
        print(distance_mat[row_ind, col_ind].sum())

```

```

10703.3

```

# hw2\_p4

February 16, 2018

```
In [1]: import numpy as np
import tensorflow as tf
%matplotlib inline
import matplotlib
import matplotlib.pyplot as plt
```

```
In [2]: from tensorflow.examples.tutorials.mnist import input_data
mnist = input_data.read_data_sets("MNIST_data/", one_hot=False)
print(mnist.train.images.shape, mnist.train.labels.shape)
```

```
Extracting MNIST_data/train-images-idx3-ubyte.gz
Extracting MNIST_data/train-labels-idx1-ubyte.gz
Extracting MNIST_data/t10k-images-idx3-ubyte.gz
Extracting MNIST_data/t10k-labels-idx1-ubyte.gz
(55000, 784) (55000,)
```

```
In [3]: def show_image(image):
    plt.gray()
    plt.imshow(image.reshape(28, 28))
    plt.show()

    # add gaussian noise to image -- mean = 0, variance = 0.1
    def add_noise(image):
        mu, sigma = 0, 0.1
        gauss = np.random.normal(mu, sigma, 28*28)
        return image + gauss
```

```
In [4]: # This is an encoder
        # 2 convolutional layers and 1 fully connected layer
        def encoder(x):
            x_image = tf.reshape(x, [-1, 28, 28, 1])

            conv1 = tf.contrib.layers.conv2d(x_image, 16, [3,3], stride=2, padding='VALID')

            conv2 = tf.contrib.layers.conv2d(conv1, 32, [3,3], stride=2, padding='VALID')

            pool2_flat = tf.reshape(conv2, [-1, 6*6*32])
```

```

        fc = tf.contrib.layers.fully_connected(pool2_flat, 100)

        return fc

# This is a decoder -- well commented!
# 1 fully connected layer, 2 convolutional layers and 1 fc layer
def decoder(x):
    fc = tf.contrib.layers.fully_connected(x, 6*6*32)
    fc_layer = tf.reshape(fc, [-1, 6, 6, 32])

    deconv1 = tf.contrib.layers.conv2d_transpose(fc_layer, 16, [3,3], stride=2, padding='V
    deconv2 = tf.contrib.layers.conv2d_transpose(deconv1, 1, [3,3], stride=2, padding='V
    deconv2_flat = tf.reshape(deconv2, [-1, 27*27])
    fc = tf.contrib.layers.fully_connected(deconv2_flat, 28*28)
    return fc

In [5]: with tf.device('/device:GPU:0'):
        # noisy image
        x = tf.placeholder(tf.float32, shape=[None, 784], name='x')
        # origin image (before adding noise)
        y = tf.placeholder(tf.float32, shape=[None, 784], name='y')

        # reconstruct images
        pred = decoder(encoder(x))
        # MSE cost function
        #compute difference between reconstructed images and imges before adding noise
        cost = tf.reduce_mean(tf.squared_difference(y, pred))
        optimizer = tf.train.AdamOptimizer(1e-4).minimize(cost)

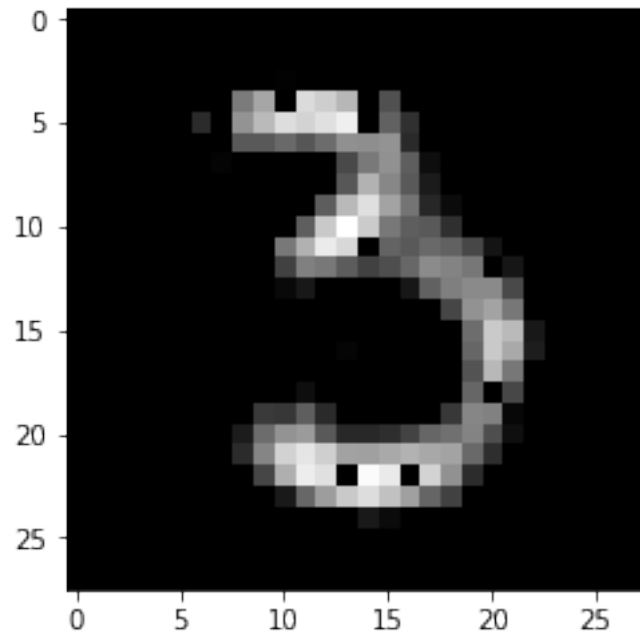
In [6]: epoch = 15000
        batch_size = 16
        config = tf.ConfigProto()
        config.gpu_options.allow_growth = True

        sess = tf.Session(config=config)
        sess.run(tf.global_variables_initializer())

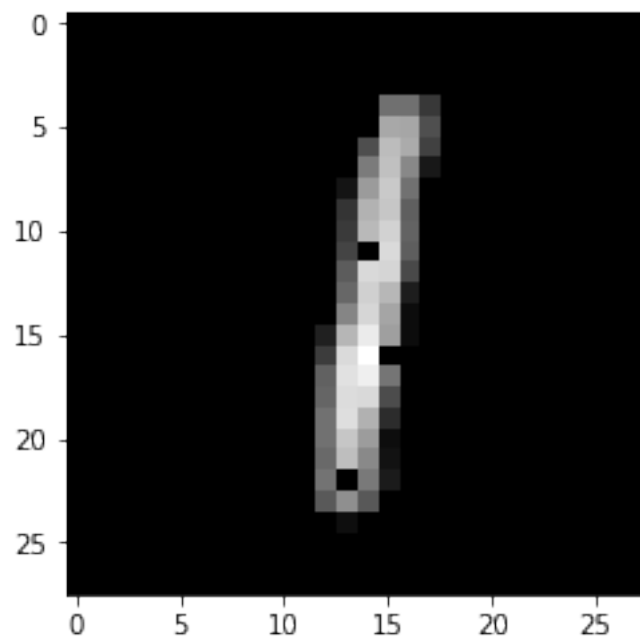
        for i in range(epoch):
            batch_x, _ = mnist.train.next_batch(batch_size)
            _, batch_cost = sess.run([optimizer, cost], feed_dict = {x:add_noise(batch_x), y:batch_x})

            # show reconstructed images after every 5000 iterations of training
            if (i+1) % 5000 == 0:
                batch_x, _ = mnist.test.next_batch(1)
                noise_image = add_noise(batch_x)
                gen_image, test_cost = sess.run([pred,cost], feed_dict = {x:noise_image, y:batch_x})
                #show_image(noise_image)
                show_image(gen_image)
                print('%d epochs-- train cost: %f    test cost: %f' % (i+1, batch_cost, test_cost))

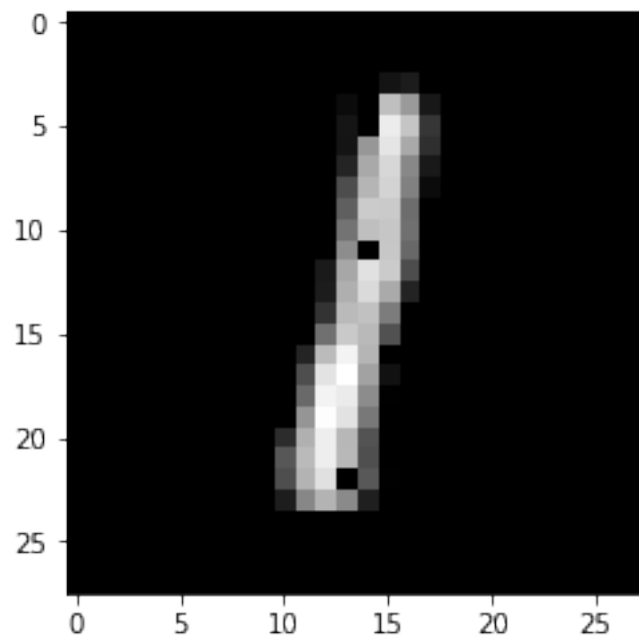
```



5000 epochs-- train cost: 0.015710    test cost: 0.017580



10000 epochs-- train cost: 0.013543      test cost: 0.005779

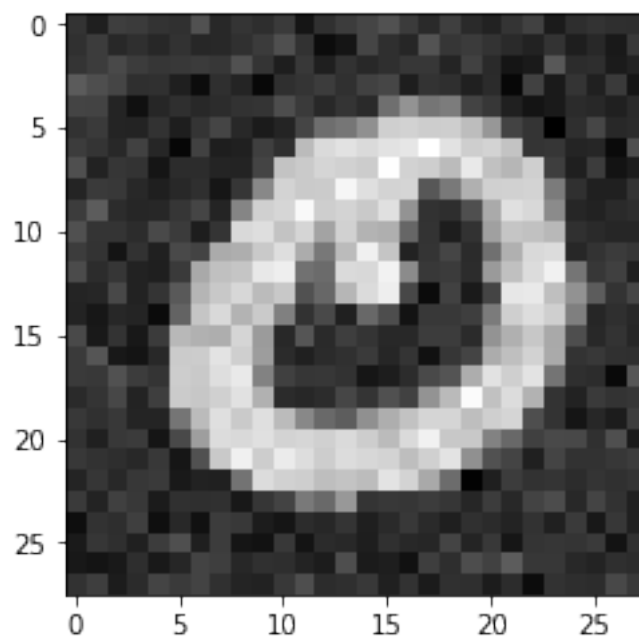


15000 epochs-- train cost: 0.014058      test cost: 0.006856

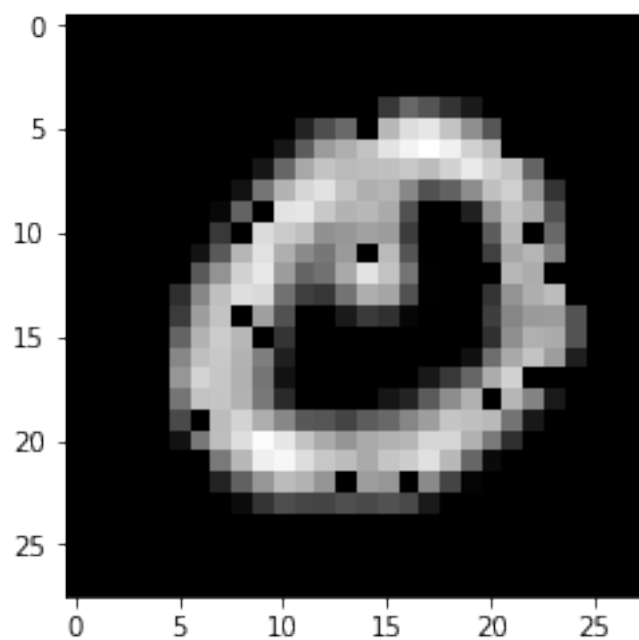
```
In [7]: # Testing -- reconstruct images from test set
        batch_x, _ = mnist.test.next_batch(1)
        noise_image = add_noise(batch_x)
        gen_image, test_cost = sess.run([pred, cost], feed_dict = {x: noise_image, y: batch_x})
        print('Noisy image')
        show_image(noise_image)
        print('Generated image')
        show_image(gen_image)
```

Noisy image



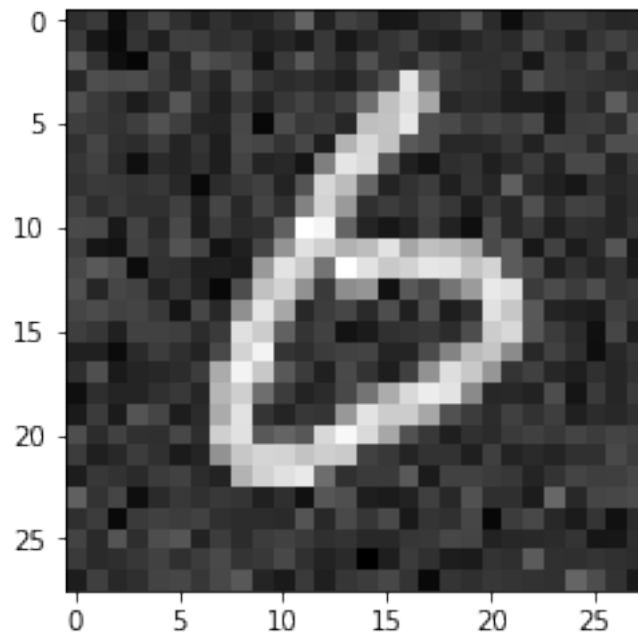


Generated image



```
In [8]: # Testing -- reconstruct images from test set
        batch_x, _ = mnist.test.next_batch(1)
        noise_image = add_noise(batch_x)
        gen_image, test_cost = sess.run([pred, cost], feed_dict = {x: noise_image, y: batch_x})
        print('Noisy image')
        show_image(noise_image)
        print('Generated image')
        show_image(gen_image)
```

Noisy image



Generated image

