```
In [1]:
        import tensorflow as tf
        import numpy as np
        import pickle as pkl
        from sklearn.manifold import TSNE
        from flip gradient import flip gradient
        from utils import *
        from timeit import default_timer as timer
        import numpy as np
        from keras import backend as K
        from keras.datasets import mnist
        from keras.layers import Conv2D,Dense,MaxPool2D,Flatten
        from keras.models import Sequential
        from sklearn.ensemble import ExtraTreesClassifier
        from sklearn.tree import DecisionTreeClassifier
        from sklearn.metrics import accuracy score
        from keras helper import NNWeightHelper
        from snes import SNES
        from sklearn.ensemble import RandomForestClassifier
        from sklearn.metrics import accuracy score
        import scipy.misc
```

/anaconda2/lib/python2.7/site-packages/h5py/\_\_init\_\_.py:36: FutureWarning: Conversi on of the second argument of issubdtype from `float` to `np.floating` is deprecate d. In future, it will be treated as `np.float64 == np.dtype(float).type`. from .\_conv import register\_converters as \_register\_converters
Using TensorFlow backend.

```
In [2]:
        from tensorflow.examples.tutorials.mnist import input data
        mnist = input_data.read_data_sets('MNIST_data', one_hot=True)
        # Process MNIST
        mnist_train = (mnist.train.images > 0).reshape(55000, 28, 28, 1).astype(np.uint8) * 25
        mnist_train = np.concatenate([mnist_train, mnist_train, mnist_train], 3)
        mnist_test = (mnist.test.images > 0).reshape(10000, 28, 28, 1).astype(np.uint8) * 255
        mnist_test = np.concatenate([mnist_test, mnist_test, mnist_test], 3)
        mnist valid = (mnist.validation.images > 0).reshape(mnist.validation.images.shape[0],2
        mnist_valid= np.concatenate([mnist_valid,mnist_valid,mnist_valid],axis=3)
        print(type(mnist_train[0]))
        # Load MNIST-M
        mnistm = pkl.load(open('mnistm data.pkl', 'rb'))
        mnistm train = mnistm['train']
        mnistm test = mnistm['test']
        mnistm valid = mnistm['valid']
        # Compute pixel mean for normalizing data
        pixel mean = np.vstack([mnist train, mnistm train]).mean((0, 1, 2))
        # Create a mixed dataset for TSNE visualization
        num\_test = 500
        combined test imgs = np.vstack([mnist test[:num test], mnistm test[:num test]])
        combined_test_labels = np.vstack([mnist.test.labels[:num_test], mnist.test.labels[:num_test]
        combined_test_domain = np.vstack([np.tile([1., 0.], [num_test, 1]),
                np.tile([0., 1.], [num_test, 1])])
```

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
<type 'numpy.ndarray'>

```
In [3]: imshow_grid(mnist_train)
  imshow_grid(mnistm_train)
```





```
In [4]:
        mnist_train_labels = mnist.train.labels
        mnist_test_labels = mnist.test.labels
        mnist valid labels = mnist.validation.labels
        print (len(mnist_train_labels), len(mnist_test_labels))
        (55000, 10000)
In [5]: # input image dimensions
        img_rows, img_cols = 28, 28
        num classes = 10
        input_shape = (img_rows, img_cols, 3)
In [6]: x_train, y_train = mnist_train , mnist_train_labels
        x_valid, y_valid = mnist_valid, mnist_valid_labels
        Mx_train, My_train = mnistm_train , mnist_train_labels
        Mx_valid, My_valid = mnistm_valid, mnist_valid_labels
In [7]: model = Sequential()
        model.add(Conv2D(32,kernel size=(5,5),activation='relu',input shape=input shape))
        model.add(MaxPool2D())
        model.add((Conv2D(48,kernel_size=(3, 3), activation='relu')))
        model.add(MaxPool2D())
        model.add(Flatten())
        model.add(Dense(20,activation='relu'))
        model.compile(loss='mse',optimizer='adam')
        model = model
        model.summary()
```

Layer (type)	Output	Shape	Param #
conv2d_1 (Conv2D)	(None,	24, 24, 32)	2432
<pre>max_pooling2d_1 (MaxPooling2</pre>	(None,	12, 12, 32)	0
conv2d_2 (Conv2D)	(None,	10, 10, 48)	13872
max_pooling2d_2 (MaxPooling2	(None,	5, 5, 48)	0
flatten_1 (Flatten)	(None,	1200)	0
dense_1 (Dense)	(None,	20)	24020
Total params: 40,324 Trainable params: 40,324 Non-trainable params: 0			

```
In [8]: nnw = NNWeightHelper(model)
         weights = nnw.get weights()
In [11]:
         SX=np.asarray(x train)
         SY=np.asarray(y_train)
         SVX=np.asarray(x valid)
         SVY=np.asarray(y valid)
         input shape=SX.shape[1:]
         TX = np.asarray(Mx train)
         TY = np.asarray(My_train)
         TVX = np.asarray(Mx valid)
         TVY = np.asarray(My_valid)
         MX=np.concatenate((SX, TX), axis=0)
         MY=np.concatenate((np.zeros(SX.shape[0]), np.ones(TX.shape[0])),axis=0)
         MVX = np.concatenate((SVX, TVX),axis=0)
         MVY = np.concatenate((np.zeros(SVX.shape[0]), np.ones(TVX.shape[0])),axis=0)
         input shape=SX.shape[1:]
In [12]: def Train classifier(x,y):
             x features=model.predict(x)
             clf=RandomForestClassifier(n estimators=18)
             clf=clf.fit(x features, y)
             return clf
         def Predict from clf(clf,x):
             x features=model.predict(x)
             y=clf.predict(x_features)
             return y
```

```
clf=RandomForestClassifier(n_estimators=18)
clf=clf.fit(x_features, y)
return clf

def Predict_from_clf(clf,x):
    x_features=model.predict(x)
    y=clf.predict(x_features)
    return y

def saveModel(filename):
    file=open(filename+'.json','w')
    file.write(model.to_json())
    model.save_weights(filename+'.h5')
    file.close()

def load_model(filename):
    file=open(filename+'.json')
    model=file.read()
    model=model_from_json(model)
    model=model_from_json(model)
    model.load_weight(filename+'.json')

def new_score_fun(label_accuracy,domain_accuray):
    temp =label_accuracy-domain_accuray
    temp /= max(domain_accuray, label_accuracy)
    return temp
```

```
In [13]: label_classifier = RandomForestClassifier()
Domain_classifier = RandomForestClassifier()
```

```
In [14]:
         RD Indices = np.random.choice(a=list(range(MX.shape[0])),size=1024)
         M D TX = MX[RD Indices]
         M D TY = MY[RD Indices]
         RS Indices = np.random.choice(a=list(range(SX.shape[0])),size=4024)
         S TX = SX[RS Indices]
         S TY = SY[RS Indices]
         RDV indices = np.random.choice(a=list(range(MVX.shape[0])), size=1024)
         M D VX = MVX[RDV indices]
         M_D_VY = MVY[RDV_indices]
         SV_indices = np.random.choice(a=list((range(SVX.shape[0]))),size=1024)
         validX = SVX[SV indices]
         validY = SVY[SV_indices]
In [15]:
         label classifier = Train classifier(SX,SY)
         label_pred = Predict_from_clf(label_classifier, validX)
         label_accuracy = accuracy_score(validY,label_pred)
         print ('label_accuracy', label_accuracy )
         Domain classifier = Train classifier(M D TX, M D TY)
         domain pred = Predict from clf(Domain classifier, M D VX)
         domain accuracy = accuracy score(M D VY, domain pred)
         print ('domain_accuracy',domain_accuracy )
         weight modifier=NNWeightHelper(model)
         weights=weight modifier.get weights()
         print 'weights to evolve:',len(weights)
         ('label accuracy', 0.3662109375)
         ('domain_accuracy', 0.8291015625)
         weights to evolve: 40324
In [16]: snes=SNES(weights,1,20)
```

```
In [17]: for i in range(60):
             start = timer()
             new weights=snes.ask()
             complied score = []
             domain accuracys = []
             label accuracys = []
             for w in new weights:
                 weight_modifier.set_weights(w)
                 label_classifier=Train_classifier(S_TX,S_TY)
                 label predictions = Predict from clf(label classifier, validX)
                 label_accuracy=accuracy_score(validY,label_predictions)
                   print ('label_accuracy',label_accuracy )
                 Domain_classifier = Train_classifier(M_D_TX,M_D_TY)
                 domain_predictions=Predict_from_clf(Domain_classifier,M_D_VX)
                 domain_accuracy=accuracy_score(M_D_VY,domain_predictions)
                   print ('domain_accuracy',domain_accuracy )
                 complied score.append(new score fun(label accuracy, domain accuracy))
                 domain accuracys.append(domain accuracy)
                 label accuracys.append(label accuracy)
             snes.tell(new weights,complied score)
             most fit model=np.argmax(complied score)
             end = timer()
             print("It took", end - start, "seconds to complete generation")
             print("the fit model has label accuracy: %0.3f and domain accuracy: %0.3f and fitne
                   %(label accuracys[most fit model], domain accuracys[most fit model], complie
         ('Step', 1, ':', -0.5269978401727862, 'best:', -0.5269978401727862, 20)
         ('It took', 40.6933331489563, 'seconds to complete generation')
         the fit model has label_accuracy: 0.428 and domain_accuracy:0.904 and fitness_scor
         e:-0.527
         ('Step', 2, ':', -0.4723404255319149, 'best:', -0.4723404255319149, 20)
         ('It took', 40.73397397994995, 'seconds to complete generation')
         the fit model has label accuracy: 0.484 and domain accuracy: 0.918 and fitness scor
         e:-0.472
         ('Step', 3, ':', -0.4851063829787234, 'best:', -0.4723404255319149, 20)
         ('It took', 36.06898593902588, 'seconds to complete generation')
         the fit model has label_accuracy: 0.473 and domain_accuracy: 0.918 and fitness_scor
         e:-0.485
         ('Step', 4, ':', -0.4630593132154006, 'best:', -0.4630593132154006, 20)
         ('It took', 36.09901690483093, 'seconds to complete generation')
         the fit model has label accuracy: 0.504 and domain accuracy: 0.938 and fitness scor
         e:-0.463
         ('Step', 5, ':', -0.44081632653061226, 'best:', -0.44081632653061226, 20)
         ('It took', 35.73264694213867, 'seconds to complete generation')
         the fit model has label_accuracy: 0.535 and domain_accuracy: 0.957 and fitness_scor
         e:-0.441
         ('Step', 6, ':', -0.4629237288135593, 'best:', -0.44081632653061226, 20)
         ('It took', 38.45079207420349, 'seconds to complete generation')
         the fit model has label_accuracy: 0.495 and domain_accuracy:0.922 and fitness_scor
         e:-0.463
         ('Step', 7, ':', -0.4647739221871714, 'best:', -0.44081632653061226, 20)
         ('It took', 38.4210479259491, 'seconds to complete generation')
         the fit model has label_accuracy: 0.497 and domain_accuracy:0.929 and fitness_scor
         e:-0.465
         ('Step', 8, ':', -0.5048025613660619, 'best:', -0.44081632653061226, 20)
         ('It took', 38.471384048461914, 'seconds to complete generation')
         the fit model has label accuracy: 0.453 and domain accuracy: 0.915 and fitness scor
         e:-0.505
         ('Step', 9, ':', -0.4309031556039173, 'best:', -0.4309031556039173, 20)
         ('It took', 38.18576383590698, 'seconds to complete generation')
```

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the fit model has label accuracy: 0.511 and domain accuracy: 0.897 and fitness scor
e:-0.431
('Step', 10, ':', -0.39397089397089397, 'best:', -0.39397089397089397, 20)
('It took', 36.95579695701599, 'seconds to complete generation')
the fit model has label accuracy: 0.569 and domain accuracy: 0.939 and fitness scor
('Step', 11, ':', -0.416666666666667, 'best:', -0.39397089397089397, 20)
('It took', 39.99688196182251, 'seconds to complete generation')
the fit model has label_accuracy: 0.540 and domain_accuracy:0.926 and fitness_scor
e:-0.417
('Step', 12, ':', -0.3935617860851506, 'best:', -0.3935617860851506, 20)
('It took', 37.453953981399536, 'seconds to complete generation')
the fit model has label_accuracy: 0.570 and domain_accuracy:0.940 and fitness_scor
e:-0.394
('Step', 13, ':', -0.37733887733887733, 'best:', -0.37733887733887733, 20)
('It took', 39.587193965911865, 'seconds to complete generation')
the fit model has label accuracy: 0.585 and domain accuracy: 0.939 and fitness scor
e:-0.377
('Step', 14, ':', -0.37563971340839303, 'best:', -0.37563971340839303, 20)
('It took', 39.667535066604614, 'seconds to complete generation')
the fit model has label_accuracy: 0.596 and domain_accuracy:0.954 and fitness scor
e:-0.376
('Step', 15, ':', -0.330208333333333, 'best:', -0.33020833333333, 20)
('It took', 38.27355885505676, 'seconds to complete generation')
the fit model has label accuracy: 0.628 and domain accuracy: 0.938 and fitness scor
e:-0.330
('Step', 16, ':', -0.32659574468085106, 'best:', -0.32659574468085106, 20)
('It took', 40.71008801460266, 'seconds to complete generation')
the fit model has label accuracy: 0.618 and domain accuracy: 0.918 and fitness scor
('Step', 17, ':', -0.30062630480167013, 'best:', -0.30062630480167013, 20)
('It took', 40.482568979263306, 'seconds to complete generation')
the fit model has label accuracy: 0.654 and domain accuracy: 0.936 and fitness scor
e:-0.301
('Step', 18, ':', -0.3316008316008316, 'best:', -0.30062630480167013, 20)
('It took', 41.086037158966064, 'seconds to complete generation')
the fit model has label accuracy: 0.628 and domain accuracy: 0.939 and fitness scor
e:-0.332
('Step', 19, ':', -0.33127572016460904, 'best:', -0.30062630480167013, 20)
('It took', 41.590157985687256, 'seconds to complete generation')
the fit model has label_accuracy: 0.635 and domain_accuracy: 0.949 and fitness_scor
e:-0.331
('Step', 20, ':', -0.335423197492163, 'best:', -0.30062630480167013, 20)
('It took', 40.2183141708374, 'seconds to complete generation')
the fit model has label_accuracy: 0.621 and domain_accuracy: 0.935 and fitness_scor
e:-0.335
('Step', 21, ':', -0.3266171792152704, 'best:', -0.30062630480167013, 20)
('It took', 39.08803200721741, 'seconds to complete generation')
the fit model has label accuracy: 0.620 and domain accuracy: 0.921 and fitness scor
e:-0.327
('Step', 22, ':', -0.3050847457627119, 'best:', -0.30062630480167013, 20)
('It took', 40.74550986289978, 'seconds to complete generation')
the fit model has label accuracy: 0.641 and domain accuracy: 0.922 and fitness scor
e:-0.305
('Step', 23, ':', -0.2861602497398543, 'best:', -0.2861602497398543, 20)
('It took', 40.2285680770874, 'seconds to complete generation')
the fit model has label accuracy: 0.670 and domain accuracy: 0.938 and fitness scor
('Step', 24, ':', -0.3044838373305527, 'best:', -0.2861602497398543, 20)
('It took', 39.85899114608765, 'seconds to complete generation')
the fit model has label_accuracy: 0.651 and domain_accuracy: 0.937 and fitness_scor
('Step', 25, ':', -0.29958677685950413, 'best:', -0.2861602497398543, 20)
('It took', 38.89574599266052, 'seconds to complete generation')
the fit model has label_accuracy: 0.662 and domain_accuracy: 0.945 and fitness_scor
('Step', 26, ':', -0.2806282722513089, 'best:', -0.2806282722513089, 20)
('It took', 41.240391969680786, 'seconds to complete generation')
the fit model has label_accuracy: 0.671 and domain_accuracy:0.933 and fitness_scor
e:-0.281
```

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('Step', 27, ':', -0.2864637985309549, 'best:', -0.2806282722513089, 20)
('It took', 38.86373496055603, 'seconds to complete generation')
the fit model has label accuracy: 0.664 and domain accuracy: 0.931 and fitness scor
e:-0.286
('Step', 28, ':', -0.2843551797040169, 'best:', -0.2806282722513089, 20)
('It took', 42.67580008506775, 'seconds to complete generation')
the fit model has label accuracy: 0.661 and domain accuracy: 0.924 and fitness scor
('Step', 29, ':', -0.2753468516542156, 'best:', -0.2753468516542156, 20)
('It took', 40.232882022857666, 'seconds to complete generation')
the fit model has label_accuracy: 0.663 and domain_accuracy:0.915 and fitness_scor
e:-0.275
('Step', 30, ':', -0.273784355179704, 'best:', -0.273784355179704, 20)
('It took', 42.03857088088989, 'seconds to complete generation')
the fit model has label accuracy: 0.671 and domain accuracy: 0.924 and fitness scor
('Step', 31, ':', -0.27415966386554624, 'best:', -0.273784355179704, 20)
('It took', 40.271496057510376, 'seconds to complete generation')
the fit model has label accuracy: 0.675 and domain accuracy: 0.930 and fitness scor
e:-0.274
('Step', 32, ':', -0.26834381551362685, 'best:', -0.26834381551362685, 20)
('It took', 41.6530499458313, 'seconds to complete generation')
the fit model has label accuracy: 0.682 and domain accuracy: 0.932 and fitness scor
e:-0.268
('Step', 33, ':', -0.2444444444444444444, 'best:', -0.2444444444444444, 20)
('It took', 38.141144037246704, 'seconds to complete generation')
the fit model has label accuracy: 0.697 and domain accuracy: 0.923 and fitness scor
('Step', 34, ':', -0.25937834941050375, 'best:', -0.244444444444444444, 20)
('It took', 43.60002088546753, 'seconds to complete generation')
the fit model has label accuracy: 0.675 and domain accuracy: 0.911 and fitness scor
('Step', 35, ':', -0.2708113804004215, 'best:', -0.24444444444444444, 20)
('It took', 41.69706583023071, 'seconds to complete generation')
the fit model has label_accuracy: 0.676 and domain_accuracy:0.927 and fitness_scor
('Step', 36, ':', -0.2611464968152866, 'best:', -0.24444444444444444, 20)
('It took', 40.41336393356323, 'seconds to complete generation')
the fit model has label_accuracy: 0.680 and domain_accuracy:0.920 and fitness_scor
e:-0.261
('It took', 41.0063841342926, 'seconds to complete generation')
the fit model has label_accuracy: 0.671 and domain_accuracy: 0.903 and fitness_scor
e:-0.257
('It took', 42.19031596183777, 'seconds to complete generation')
the fit model has label_accuracy: 0.693 and domain_accuracy: 0.938 and fitness_scor
e:-0.260
('Step', 39, ':', -0.20613107822410148, 'best:', -0.20613107822410148, 20)
('It took', 41.03705310821533, 'seconds to complete generation')
the fit model has label_accuracy: 0.733 and domain_accuracy: 0.924 and fitness_scor
e:-0.206
('Step', 40, ':', -0.2601880877742947, 'best:', -0.20613107822410148, 20)
('It took', 43.284093141555786, 'seconds to complete generation')
the fit model has label accuracy: 0.691 and domain accuracy: 0.935 and fitness scor
e:-0.260
('Step', 41, ':', -0.24572649572649571, 'best:', -0.20613107822410148, 20)
('It took', 39.87771987915039, 'seconds to complete generation')
the fit model has label accuracy: 0.689 and domain accuracy: 0.914 and fitness scor
e:-0.246
('Step', 42, ':', -0.2603613177470776, 'best:', -0.20613107822410148, 20)
('It took', 41.78398895263672, 'seconds to complete generation')
the fit model has label_accuracy: 0.680 and domain_accuracy: 0.919 and fitness_scor
e:-0.260
('Step', 43, ':', -0.20979765708200213, 'best:', -0.20613107822410148, 20)
('It took', 39.65928912162781, 'seconds to complete generation')
the fit model has label_accuracy: 0.725 and domain_accuracy: 0.917 and fitness_scor
e:-0.210
('Step', 44, ':', -0.23435843054082714, 'best:', -0.20613107822410148, 20)
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('It took', 39.42806696891785, 'seconds to complete generation')
the fit model has label accuracy: 0.705 and domain accuracy: 0.921 and fitness scor
e:-0.234
('Step', 45, ':', -0.21264994547437296, 'best:', -0.20613107822410148, 20)
('It took', 41.2503879070282, 'seconds to complete generation')
the fit model has label accuracy: 0.705 and domain accuracy: 0.896 and fitness scor
e:-0.213
('Step', 46, ':', -0.208067940552017, 'best:', -0.20613107822410148, 20)
('It took', 40.72658705711365, 'seconds to complete generation')
the fit model has label_accuracy: 0.729 and domain_accuracy: 0.920 and fitness_scor
e:-0.208
('Step', 47, ':', -0.20276008492569003, 'best:', -0.20276008492569003, 20)
('It took', 39.3925940990448, 'seconds to complete generation')
the fit model has label_accuracy: 0.733 and domain_accuracy: 0.920 and fitness_scor
e:-0.203
('Step', 48, ':', -0.20191285866099895, 'best:', -0.20191285866099895, 20)
('It took', 39.929206132888794, 'seconds to complete generation')
the fit model has label accuracy: 0.733 and domain accuracy: 0.919 and fitness scor
e:-0.202
('Step', 49, ':', -0.20234291799787008, 'best:', -0.20191285866099895, 20)
('It took', 41.22449803352356, 'seconds to complete generation')
the fit model has label accuracy: 0.731 and domain accuracy: 0.917 and fitness scor
e:-0.202
('Step', 50, ':', -0.17282608695652174, 'best:', -0.17282608695652174, 20)
('It took', 40.8322970867157, 'seconds to complete generation')
the fit model has label accuracy: 0.743 and domain accuracy: 0.898 and fitness scor
e:-0.173
('Step', 51, ':', -0.18763326226012794, 'best:', -0.17282608695652174, 20)
('It took', 40.65230393409729, 'seconds to complete generation')
the fit model has label accuracy: 0.744 and domain accuracy: 0.916 and fitness scor
e:-0.188
('Step', 52, ':', -0.1819148936170213, 'best:', -0.17282608695652174, 20)
('It took', 39.593459129333496, 'seconds to complete generation')
the fit model has label_accuracy: 0.751 and domain_accuracy: 0.918 and fitness_scor
e:-0.182
('Step', 53, ':', -0.19894736842105262, 'best:', -0.17282608695652174, 20)
('It took', 41.122143030166626, 'seconds to complete generation')
the fit model has label accuracy: 0.743 and domain accuracy: 0.928 and fitness scor
e:-0.199
('Step', 54, ':', -0.19135135135135135134, 'best:', -0.17282608695652174, 20)
('It took', 43.967583894729614, 'seconds to complete generation')
the fit model has label_accuracy: 0.730 and domain_accuracy:0.903 and fitness_scor
e:-0.191
('Step', 55, ':', -0.18619246861924685, 'best:', -0.17282608695652174, 20)
('It took', 40.919296979904175, 'seconds to complete generation')
the fit model has label_accuracy: 0.760 and domain_accuracy: 0.934 and fitness_scor
e:-0.186
('Step', 56, ':', -0.1810344827586207, 'best:', -0.17282608695652174, 20)
('It took', 40.91202688217163, 'seconds to complete generation')
the fit model has label accuracy: 0.742 and domain accuracy: 0.906 and fitness scor
e:-0.181
('Step', 57, ':', -0.16648648648648648, 'best:', -0.16648648648648, 20)
('It took', 41.41397213935852, 'seconds to complete generation')
the fit model has label accuracy: 0.753 and domain accuracy: 0.903 and fitness scor
e:-0.166
('Step', 58, ':', -0.17326203208556148, 'best:', -0.16648648648648648, 20)
('It took', 42.29031991958618, 'seconds to complete generation')
the fit model has label accuracy: 0.755 and domain accuracy: 0.913 and fitness scor
e:-0.173
('Step', 59, ':', -0.17880085653104924, 'best:', -0.16648648648648648, 20)
('It took', 39.26325798034668, 'seconds to complete generation')
the fit model has label accuracy: 0.749 and domain accuracy: 0.912 and fitness scor
e:-0.179
('Step', 60, ':', -0.17330462863293863, 'best:', -0.16648648648648648, 20)
('It took', 40.730553150177, 'seconds to complete generation')
the fit model has label_accuracy: 0.750 and domain_accuracy: 0.907 and fitness_scor
e:-0.173
```

```
In [19]: weight modifier.set weights(snes.center)
In [31]: label classifier = Train classifier(mnist train,mnist train labels)
         sourcePredictions = Predict from clf(label classifier, mnist test)
         sourceAccuray = accuracy_score(mnist_test_labels,sourcePredictions)
         print "label predicitions on MNIST %0.3f" % (sourceAccuray)
         targetPredictions = Predict from clf(label classifier, mnistm test)
         targetAccuracy = accuracy_score(targetPredictions, mnist test labels)
         print "label predicitons on MNISTM %0.3f" %(targetAccuracy)
         label predicitions on MNIST 0.838
         label predicitons on MNISTM 0.163
In [32]: # mixing source and target data
         R_MT_INDEX = np.random.choice(a=list(range(MX.shape[0])), size=MY.shape[0])
         TM X = MX[R MT INDEX]
         TM Y = MY[R MT INDEX]
         TEST MX = np.concatenate([mnist test, mnistm test], axis=0)
         TEST MY = np.concatenate([np.zeros(mnist test.shape[0]), np.ones(mnistm test.shape[0])
In [33]:
         Domain classifier = Train classifier(TM X, TM Y)
         domain pred = Predict from clf(Domain classifier, TEST MX)
         domain accuracy = accuracy score(TEST MY, domain pred)
         print("domain predicitions accuracy: %0.3f" %(domain accuracy))
         domain predicitions accuracy: 0.946
         import pandas as pd
In [23]:
In [29]: log = pd.read csv('log.csv', header=None)
In [30]: log.plot(x=0, y=1, style='o')
Out[30]: <matplotlib.axes. subplots.AxesSubplot at 0x1c32d0a990>
          -0.15
                   1
          -0.20
          -0.25
          -0.30
          -0.35
          -0.40
          -0.45
```

50

60

-0.50

30