```
In [52]:
         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, Dropout
         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
```

```
In [53]:
         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 [55]: imshow_grid(mnist_train)
 imshow_grid(mnistm_train)





```
In [56]: 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))
# print (mnist_train_labels[0:4])
```

(55000, 10000)

```
In [57]: # input image dimensions
  img_rows, img_cols = 28, 28
  num_classes = 10
  input_shape = (img_rows, img_cols, 3)
```

```
In [75]: 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 [76]: 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(Dropout(0.1))
  model.add(Dropout(0.1))
  model.add(Dense(10,activation='relu'))
  model.compile(loss='mse',optimizer='adam')
  model = model
  model.summary()
```

Layer (type)	Output	Shape	Param #
conv2d_5 (Conv2D)	(None,	24, 24, 32)	2432
max_pooling2d_5 (MaxPooling2	(None,	12, 12, 32)	0
conv2d_6 (Conv2D)	(None,	10, 10, 48)	13872
max_pooling2d_6 (MaxPooling2	(None,	5, 5, 48)	0
flatten_3 (Flatten)	(None,	1200)	0
dropout_3 (Dropout)	(None,	1200)	0
dense_3 (Dense)	(None,	10)	12010
Total params: 28,314 Trainable params: 28,314 Non-trainable params: 0			

```
In [77]: nnw = NNWeightHelper(model)
weights = nnw.get_weights()
```

```
In [78]: SX=np.asarray(x_train)
    SY=np.asarray(y_train)
    SVX=np.asarray(y_valid)
    SVY=np.asarray(y_valid)
    input_shape=SX.shape[1:]

IX = np.asarray(Mx_train)
    TY = np.asarray(My_train)
    TVX = np.asarray(Mx_valid)
    TVY = np.asarray(My_valid)

IVY = 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 [79]: 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
         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.load_weight(filename+'.json')
         def new score fun(label accuracy, domain accuray):
             return label accuracy - (4*(np.square(domain accuray-0.5)))
In [80]:
         label clf = RandomForestClassifier()
         domain clf = RandomForestClassifier()
         RD Indices = np.random.choice(a=list(range(MX.shape[0])),size=1024)
In [81]:
         M D TX = MX[RD Indices]
         M D TY = MY[RD Indices]
         RS Indices = np.random.choice(a=list(range(SX.shape[0])),size=1024)
         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 [82]:
         label clf = Train classifier(SX,SY)
         label pred = Predict from clf(label clf, validX)
         label accuracy = accuracy score(validY, label pred)
         print ('label accuracy', label accuracy )
         domain clf = Train classifier(M D TX, M D TY)
         domain_pred = Predict_from_clf(domain_clf, 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', 0.380859375)
         ('domain_accuracy', 0.9033203125)
         weights to evolve: 28314
In [83]: | snes=SNES(weights,1,20)
```

```
In [84]: logscore = []
         for i in range(60):
             start = timer()
             new weights=snes.ask()
             complied score = []
             domain accuracys = []
             label accuracys = []
             for w in new_weights:
                 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=1024)
                 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]
                 weight modifier.set weights(w)
                 label clf=Train classifier(S TX,S TY)
                 label predictions = Predict from clf(label clf, validX)
                 label accuracy=accuracy score(validY, label predictions)
                   print ('label accuracy',label accuracy )
                 domain clf = Train classifier(M D TX,M D TY)
                 domain predictions=Predict from clf(domain clf,M D VX)
                 domain accuracy=accuracy score(M D VY,domain predictions)
                   print ('domain_accuracy',domain_accuracy )
                 new score = new score fun(label accuracy, domain accuracy)
                 complied score.append(new score)
                 domain_accuracys.append(domain_accuracy)
                 label_accuracys.append(label_accuracy)
             snes.tell(new_weights,complied_score)
             max_index=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 compl
                    %(label_accuracys[max_index], domain_accuracys[max_index], complied_score[max_index]
             logscore.append(complied_score[max_index])
         ('Step', 1, ':', -0.07291030883789062, 'best:', -0.07291030883789062, 20)
         ('It took', 20.770617961883545, 'seconds to complete generation')
         the fit model has label accuracy: 0.203 and domain accuracy: 0.763 and fitness scor
         ('Step', 2, ':', 0.0, 'best:', 0.0, 20)
         ('It took', 20.403969049453735, 'seconds to complete generation')
         the fit model has label_accuracy: 0.165 and domain_accuracy: 0.703 and fitness_scor
         e:0.000
         ('Step', 3, ':', -0.027744293212890625, 'best:', 0.0, 20)
         ('It took', 20.43546199798584, 'seconds to complete generation')
         the fit model has label_accuracy: 0.187 and domain_accuracy:0.731 and fitness_scor
         e:-0.028
         ('Step', 4, ':', -0.052677154541015625, 'best:', 0.0, 20)
         ('It took', 20.26436710357666, 'seconds to complete generation')
         the fit model has label accuracy: 0.279 and domain accuracy: 0.788 and fitness scor
         e:-0.053
         ('Step', 5, ':', 0.00726318359375, 'best:', 0.00726318359375, 20)
```

```
('It took', 20.314198970794678, 'seconds to complete generation')
the fit model has label accuracy: 0.179 and domain accuracy: 0.707 and fitness scor
e:0.007
('Step', 6, ':', -0.014163970947265625, 'best:', 0.00726318359375, 20)
('It took', 20.225824117660522, 'seconds to complete generation')
the fit model has label accuracy: 0.128 and domain accuracy: 0.688 and fitness scor
e:-0.014
('Step', 7, ':', 0.014492034912109375, 'best:', 0.014492034912109375, 20)
('It took', 20.37145495414734, 'seconds to complete generation')
the fit model has label_accuracy: 0.137 and domain_accuracy:0.675 and fitness_scor
e:0.014
('Step', 8, ':', 0.07525253295898438, 'best:', 0.07525253295898438, 20)
('It took', 20.386691093444824, 'seconds to complete generation')
the fit model has label_accuracy: 0.182 and domain_accuracy: 0.663 and fitness_scor
e:0.075
('Step', 9, ':', 0.072021484375, 'best:', 0.07525253295898438, 20)
('It took', 20.438518047332764, 'seconds to complete generation')
the fit model has label accuracy: 0.113 and domain accuracy: 0.602 and fitness scor
e:0.072
('Step', 10, ':', 0.09954833984375, 'best:', 0.09954833984375, 20)
('It took', 20.06223702430725, 'seconds to complete generation')
the fit model has label_accuracy: 0.132 and domain_accuracy:0.590 and fitness scor
e:0.100
('Step', 11, ':', 0.0976409912109375, 'best:', 0.09954833984375, 20)
('It took', 20.25060200691223, 'seconds to complete generation')
the fit model has label accuracy: 0.158 and domain accuracy: 0.623 and fitness scor
e:0.098
('Step', 12, ':', 0.13271713256835938, 'best:', 0.13271713256835938, 20)
('It took', 19.945361137390137, 'seconds to complete generation')
the fit model has label accuracy: 0.200 and domain accuracy: 0.630 and fitness scor
e:0.133
('Step', 13, ':', 0.08203125, 'best:', 0.13271713256835938, 20)
('It took', 19.81911301612854, 'seconds to complete generation')
the fit model has label_accuracy: 0.130 and domain_accuracy: 0.609 and fitness_scor
e:0.082
('Step', 14, ':', 0.12142562866210938, 'best:', 0.13271713256835938, 20)
('It took', 19.930845022201538, 'seconds to complete generation')
the fit model has label accuracy: 0.172 and domain accuracy: 0.612 and fitness scor
e:0.121
('Step', 15, ':', 0.0779876708984375, 'best:', 0.13271713256835938, 20)
('It took', 19.98439598083496, 'seconds to complete generation')
the fit model has label_accuracy: 0.097 and domain_accuracy: 0.568 and fitness_scor
e:0.078
('Step', 16, ':', 0.13570785522460938, 'best:', 0.13570785522460938, 20)
('It took', 19.976675033569336, 'seconds to complete generation')
the fit model has label_accuracy: 0.195 and domain_accuracy: 0.622 and fitness_scor
e:0.136
('Step', 17, ':', 0.12084579467773438, 'best:', 0.13570785522460938, 20)
('It took', 19.97342300415039, 'seconds to complete generation')
the fit model has label accuracy: 0.155 and domain accuracy: 0.593 and fitness scor
e:0.121
('Step', 18, ':', 0.17620849609375, 'best:', 0.17620849609375, 20)
('It took', 20.021899938583374, 'seconds to complete generation')
the fit model has label accuracy: 0.251 and domain accuracy: 0.637 and fitness scor
e:0.176
('Step', 19, ':', 0.07369613647460938, 'best:', 0.17620849609375, 20)
('It took', 20.21200203895569, 'seconds to complete generation')
the fit model has label accuracy: 0.091 and domain accuracy: 0.565 and fitness scor
e:0.074
('Step', 20, ':', 0.08294677734375, 'best:', 0.17620849609375, 20)
('It took', 20.157356023788452, 'seconds to complete generation')
the fit model has label accuracy: 0.186 and domain accuracy: 0.660 and fitness scor
e:0.083
('Step', 21, ':', 0.0830078125, 'best:', 0.17620849609375, 20)
('It took', 19.800989866256714, 'seconds to complete generation')
the fit model has label_accuracy: 0.118 and domain_accuracy: 0.594 and fitness_scor
e:0.083
('Step', 22, ':', 0.10851669311523438, 'best:', 0.17620849609375, 20)
('It took', 20.140537977218628, 'seconds to complete generation')
the fit model has label_accuracy: 0.157 and domain_accuracy: 0.610 and fitness_scor
```

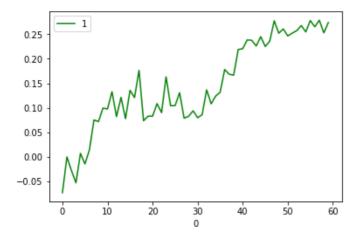
```
e:0.109
('Step', 23, ':', 0.09014511108398438, 'best:', 0.17620849609375, 20)
('It took', 19.927549123764038, 'seconds to complete generation')
the fit model has label accuracy: 0.221 and domain accuracy: 0.681 and fitness scor
e:0.090
('Step', 24, ':', 0.16308212280273438, 'best:', 0.17620849609375, 20)
('It took', 19.84715485572815, 'seconds to complete generation')
the fit model has label accuracy: 0.225 and domain accuracy: 0.624 and fitness scor
e:0.163
('Step', 25, ':', 0.10424423217773438, 'best:', 0.17620849609375, 20)
('It took', 19.836089849472046, 'seconds to complete generation')
the fit model has label_accuracy: 0.139 and domain_accuracy:0.593 and fitness_scor
e:0.104
('Step', 26, ':', 0.1044921875, 'best:', 0.17620849609375, 20)
('It took', 19.879451036453247, 'seconds to complete generation')
the fit model has label accuracy: 0.113 and domain accuracy: 0.547 and fitness scor
e:0.104
('Step', 27, ':', 0.13076400756835938, 'best:', 0.17620849609375, 20)
('It took', 19.908148050308228, 'seconds to complete generation')
the fit model has label accuracy: 0.198 and domain accuracy: 0.630 and fitness scor
e:0.131
('Step', 28, ':', 0.078857421875, 'best:', 0.17620849609375, 20)
('It took', 19.970968008041382, 'seconds to complete generation')
the fit model has label accuracy: 0.120 and domain accuracy: 0.602 and fitness scor
e:0.079
('Step', 29, ':', 0.0827484130859375, 'best:', 0.17620849609375, 20)
('It took', 19.952478170394897, 'seconds to complete generation')
the fit model has label accuracy: 0.164 and domain accuracy: 0.643 and fitness scor
e:0.083
('Step', 30, ':', 0.0939788818359375, 'best:', 0.17620849609375, 20)
('It took', 19.851691961288452, 'seconds to complete generation')
the fit model has label accuracy: 0.117 and domain accuracy: 0.576 and fitness scor
e:0.094
('Step', 31, ':', 0.079833984375, 'best:', 0.17620849609375, 20)
('It took', 19.661595106124878, 'seconds to complete generation')
the fit model has label_accuracy: 0.135 and domain_accuracy:0.617 and fitness scor
e:0.080
('Step', 32, ':', 0.08599472045898438, 'best:', 0.17620849609375, 20)
('It took', 19.905591011047363, 'seconds to complete generation')
the fit model has label_accuracy: 0.116 and domain_accuracy: 0.587 and fitness_scor
e:0.086
('Step', 33, ':', 0.13665771484375, 'best:', 0.17620849609375, 20)
('It took', 19.790705919265747, 'seconds to complete generation')
the fit model has label_accuracy: 0.169 and domain_accuracy: 0.590 and fitness_scor
e:0.137
('Step', 34, ':', 0.108154296875, 'best:', 0.17620849609375, 20)
('It took', 19.80703592300415, 'seconds to complete generation')
the fit model has label_accuracy: 0.149 and domain_accuracy: 0.602 and fitness_scor
e:0.108
('Step', 35, ':', 0.1240081787109375, 'best:', 0.17620849609375, 20)
('It took', 20.92529797554016, 'seconds to complete generation')
the fit model has label accuracy: 0.185 and domain accuracy: 0.623 and fitness scor
e:0.124
('Step', 36, ':', 0.13180160522460938, 'best:', 0.17620849609375, 20)
('It took', 20.28514289855957, 'seconds to complete generation')
the fit model has label accuracy: 0.132 and domain accuracy: 0.503 and fitness scor
e:0.132
('Step', 37, ':', 0.17809677124023438, 'best:', 0.17809677124023438, 20)
('It took', 20.27926802635193, 'seconds to complete generation')
the fit model has label_accuracy: 0.187 and domain_accuracy: 0.546 and fitness_scor
e:0.178
('Step', 38, ':', 0.16845321655273438, 'best:', 0.17809677124023438, 20)
('It took', 20.213808059692383, 'seconds to complete generation')
the fit model has label accuracy: 0.185 and domain accuracy: 0.563 and fitness scor
('Step', 39, ':', 0.16650009155273438, 'best:', 0.17809677124023438, 20)
('It took', 20.45617389678955, 'seconds to complete generation')
the fit model has label_accuracy: 0.182 and domain_accuracy:0.562 and fitness_scor
e:0.167
```

```
('Step', 40, ':', 0.21883773803710938, 'best:', 0.21883773803710938, 20)
('It took', 20.444185972213745, 'seconds to complete generation')
the fit model has label accuracy: 0.245 and domain accuracy: 0.581 and fitness scor
e:0.219
('Step', 41, ':', 0.22048568725585938, 'best:', 0.22048568725585938, 20)
('It took', 20.34163188934326, 'seconds to complete generation')
the fit model has label accuracy: 0.228 and domain accuracy: 0.542 and fitness scor
('Step', 42, ':', 0.2383880615234375, 'best:', 0.2383880615234375, 20)
('It took', 20.498934984207153, 'seconds to complete generation')
the fit model has label_accuracy: 0.240 and domain_accuracy:0.521 and fitness_scor
e:0.238
('Step', 43, ':', 0.23760604858398438, 'best:', 0.2383880615234375, 20)
('It took', 20.397908926010132, 'seconds to complete generation')
the fit model has label accuracy: 0.257 and domain accuracy: 0.569 and fitness scor
('Step', 44, ':', 0.22603988647460938, 'best:', 0.2383880615234375, 20)
('It took', 20.314714193344116, 'seconds to complete generation')
the fit model has label accuracy: 0.243 and domain accuracy: 0.565 and fitness scor
('Step', 45, ':', 0.24505615234375, 'best:', 0.24505615234375, 20)
('It took', 20.86320209503174, 'seconds to complete generation')
the fit model has label_accuracy: 0.277 and domain_accuracy:0.590 and fitness scor
('Step', 46, ':', 0.22497177124023438, 'best:', 0.24505615234375, 20)
('It took', 20.409596920013428, 'seconds to complete generation')
the fit model has label accuracy: 0.250 and domain accuracy: 0.579 and fitness scor
('Step', 47, ':', 0.23580551147460938, 'best:', 0.24505615234375, 20)
('It took', 20.272320985794067, 'seconds to complete generation')
the fit model has label accuracy: 0.250 and domain accuracy: 0.560 and fitness scor
('Step', 48, ':', 0.2774925231933594, 'best:', 0.2774925231933594, 20)
('It took', 20.3313729763031, 'seconds to complete generation')
the fit model has label accuracy: 0.280 and domain accuracy: 0.526 and fitness scor
('Step', 49, ':', 0.251953125, 'best:', 0.2774925231933594, 20)
('It took', 20.464900970458984, 'seconds to complete generation')
the fit model has label_accuracy: 0.268 and domain_accuracy: 0.562 and fitness_scor
('Step', 50, ':', 0.26068115234375, 'best:', 0.2774925231933594, 20)
('It took', 21.024746894836426, 'seconds to complete generation')
the fit model has label_accuracy: 0.274 and domain_accuracy:0.559 and fitness scor
e:0.261
('Step', 51, ':', 0.2464447021484375, 'best:', 0.2774925231933594, 20)
('It took', 21.380491018295288, 'seconds to complete generation')
the fit model has label_accuracy: 0.286 and domain_accuracy:0.600 and fitness scor
e:0.246
('Step', 52, ':', 0.2522544860839844, 'best:', 0.2774925231933594, 20)
('It took', 20.474724054336548, 'seconds to complete generation')
the fit model has label accuracy: 0.265 and domain accuracy: 0.556 and fitness scor
e:0.252
('Step', 53, ':', 0.2572288513183594, 'best:', 0.2774925231933594, 20)
('It took', 20.24232316017151, 'seconds to complete generation')
the fit model has label accuracy: 0.271 and domain accuracy: 0.558 and fitness scor
('Step', 54, ':', 0.2676658630371094, 'best:', 0.2774925231933594, 20)
('It took', 20.459728002548218, 'seconds to complete generation')
the fit model has label accuracy: 0.275 and domain accuracy: 0.544 and fitness scor
('Step', 55, ':', 0.254638671875, 'best:', 0.2774925231933594, 20)
('It took', 23.335206985473633, 'seconds to complete generation')
the fit model has label accuracy: 0.284 and domain accuracy: 0.586 and fitness scor
('Step', 56, ':', 0.2778282165527344, 'best:', 0.2778282165527344, 20)
('It took', 21.465917825698853, 'seconds to complete generation')
the fit model has label_accuracy: 0.293 and domain_accuracy: 0.562 and fitness_scor
('Step', 57, ':', 0.2650718688964844, 'best:', 0.2778282165527344, 20)
('It took', 20.7005398273468, 'seconds to complete generation')
```

```
the fit model has label accuracy: 0.271 and domain accuracy: 0.540 and fitness scor
         e:0.265
         ('Step', 58, ':', 0.2784996032714844, 'best:', 0.2784996032714844, 20)
         ('It took', 20.581918954849243, 'seconds to complete generation')
         the fit model has label accuracy: 0.290 and domain accuracy: 0.554 and fitness scor
         ('Step', 59, ':', 0.2527732849121094, 'best:', 0.2784996032714844, 20)
         ('It took', 20.369581937789917, 'seconds to complete generation')
         the fit model has label_accuracy: 0.275 and domain_accuracy: 0.575 and fitness_scor
         e:0.253
         ('Step', 60, ':', 0.2737884521484375, 'best:', 0.2784996032714844, 20)
         ('It took', 20.464088916778564, 'seconds to complete generation')
         the fit model has label accuracy: 0.305 and domain accuracy: 0.588 and fitness scor
         e:0.274
In [85]:
         saveModel('mnistmodeltest dense128 new')
         import pandas as pd
         scorelog = pd.DataFrame(logscore)
         scorelog.to csv('logscorel.csv', header=None)
In [86]: weight modifier.set weights(snes.center)
In [87]: label clf = Train classifier(mnist train,mnist train labels)
         sourcePredictions = Predict from clf(label clf, mnist test)
         sourceAccuray = accuracy score(mnist test labels, sourcePredictions)
         print "label predicitions on MNIST %0.3f" % (sourceAccuray)
         targetPredictions = Predict_from_clf(label_clf, mnistm_test)
         targetAccuracy = accuracy_score(targetPredictions, mnist_test_labels)
         print "label predicitons on MNISTM %0.3f" %(targetAccuracy)
         label predicitions on MNIST 0.268
         label predicitons on MNISTM 0.103
         R_MT_INDEX = np.random.choice(a=list(range(MX.shape[0])), size=MY.shape[0])
In [88]:
         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 [89]:
         domain clf = Train classifier(TM X, TM Y)
         domain pred = Predict from clf(domain clf, TEST MX)
         domain accuracy = accuracy score(TEST MY, domain pred)
         print("domain predicitions accuracy: %0.3f" %(domain accuracy))
         domain predicitions accuracy: 0.555
In [92]: log = pd.read csv('logscore1.csv', header=None)
         # print log
```

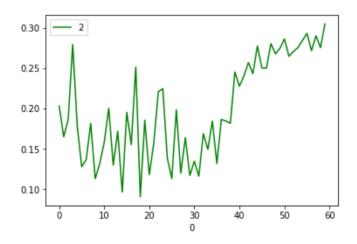
```
In [97]: # log.plot()
log.plot(x=0, y=1, style='g')
```

Out[97]: <matplotlib.axes. subplots.AxesSubplot at 0x1c247f94d0>



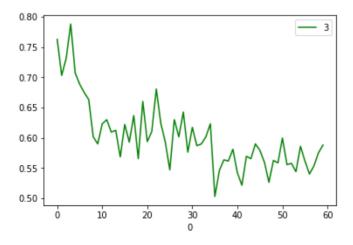
```
In [98]: log.plot(x=0, y=2, style='g')
```

Out[98]: <matplotlib.axes._subplots.AxesSubplot at 0x1c23260490>



```
In [99]: log.plot(x=0, y=3, style='g')
```

Out[99]: <matplotlib.axes._subplots.AxesSubplot at 0x1c23940550>



```
In [ ]:
```