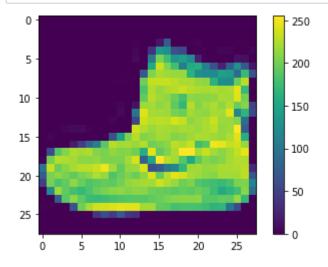
Thasina Tabashum

```
In [70]:
         from __future__ import absolute_import, division, print_function, unicode_litera
         import tensorflow as tf
         from tensorflow import keras
         # Helper libraries
         import numpy as np
         import matplotlib.pyplot as plt
         import numpy as np
         import pylab as py
         %matplotlib inline
In [71]: print(tf.__version__)
         2.1.0
         from sklearn.datasets import load digits
In [72]:
         fashion mnist = keras.datasets.fashion mnist
         (train_images, train_labels), (test_images, test_labels) = fashion_mnist.load_da
In [78]:
         print(train images.shape)
         print(train labels.shape)
         print(len(train_images))
         (60000, 28, 28)
         (60000,)
         60000
         print(test images.shape)
In [94]:
         print(test labels.shape)
         print(len(test_images))
         (10000, 28, 28)
         (10000,)
         10000
In [95]: test_images2 = test_images.reshape(len(test_images),-1)
In [79]: | train images2 = train images.reshape(len(train images),-1)
In [80]:
         print(train images2.shape)
         (60000, 784)
```

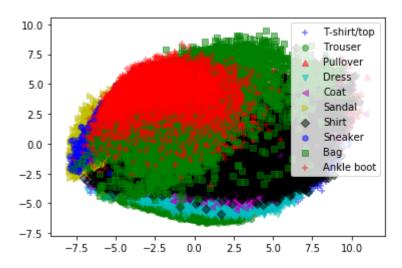
```
In [82]: plt.figure()
    plt.imshow(train_images[0])
    plt.colorbar()
    plt.grid(False)
    plt.show()
```



```
In [83]: train_images2 = train_images2 / 255.0
test_images = test_images / 255.0
```

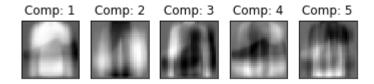


(60000, 5)



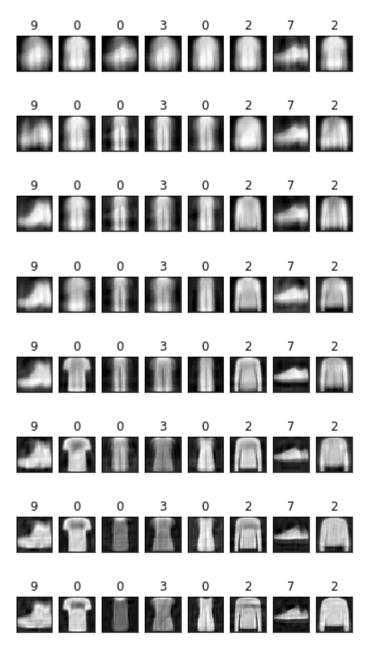
```
In [87]: def plot_gallery(data, labels, shape, interpolation='nearest'):
    for i in range(data.shape[0]):
        py.subplot(1, data.shape[0], (i + 1))
        py.imshow(data[i].reshape(shape), interpolation=interpolation)
        py.title(labels[i])
        py.xticks(()), py.yticks(())
        py.gray()
```

```
In [89]: labels = ['Comp: %d' % (i+1) for i in range(len(pca.components_)) ]
plot_gallery(pca.components_, labels, shape=(28,28))
```



```
In [93]: n = 8 # number of digits for demonstration
    dims = [1,2,3,5,10,20,40,64]
    print('compressed images of first',n,'digits')
    print('with this many PCA components:',dims)
    for d in dims: # dimensionality for compressed signal
        pca = PCA(n_components=d)
        pca.fit_transform(train_images2)
        reduced_X = pca.transform(train_images2[0:n]) # the reduced dimensionality
        recovered_X = pca.inverse_transform(reduced_X)
        py.figure()
        plot_gallery(recovered_X, train_labels[0:n], shape=(28, 28))
```

compressed images of first 8 digits with this many PCA components: [1, 2, 3, 5, 10, 20, 40, 64]



```
In [96]:
          from sklearn.naive bayes import GaussianNB
          model = GaussianNB().fit(train_images2, train_labels)
          train score = model.score(train images2, train labels)
          print('training score (overfitting!):',train score)
          test_score = model.score(test_images2, test_labels)
          print('test score:',test_score)
          training score (overfitting!): 0.5877833333333333
          test score: 0.5856
In [100]:
          pca = PCA(n components=10)
          pca.fit(train_images2)
          tX train = pca.transform(train images2)
          tX test = pca.transform(test images2)
          model = GaussianNB().fit(tX train, train labels)
          train score = model.score(tX train, train labels)
          print('training score (overfitting!):',train_score)
          test score = model.score(tX test, test labels)
          print('test score:',test_score)
          from sklearn import metrics
          y_test_pred = model.predict(tX_test)
          expected = test_labels
          predicted = model.predict(tX test)
          print("Confusion matrix:\n%s" % metrics.confusion_matrix(expected, predicted))
          training score (overfitting!): 0.70608333333333333
          test score: 0.7003
          Confusion matrix:
          [[763
                  0
                    16 75
                             12 44 61
                                           0
                                              29
                                                   0]
           [ 27 882 15
                         52
                               7
                                   6
                                       2
                                           0
                                               9
                                                   0]
             10
                  0 514
                           7 226
                                  38 175
                                           0
                                              30
                                                   0]
           [ 79
                  7
                      4 761
                             45
                                  27
                                      72
                                               5
                                                   01
                  3 257
                         42 569
                                  22
                                      93
                                              13
                                                   0]
              1
                                           0
                                              24
              0
                  0
                      0
                           2
                               0 692
                                       1 205
                                                  76]
           [236]
                  0 180 37 165
                                  76 256
                                           0
                                              50
                                                   0]
              0
                  0
                     0
                           0
                               0
                                  97
                                       0 813
                                               3
                                                  87]
              6
                  1 13
                           8
                             14
                                  53
                                      33
                                          18 852
                                                   2]
              0
                      0
                           0
                               0
                                  49
                                       2
                                          45
                                               3 901]]
```

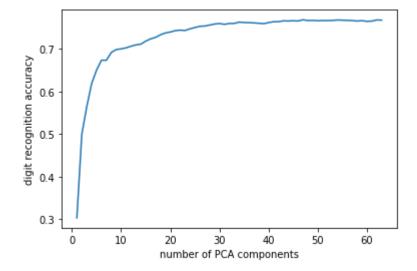
```
In [101]: accuracy = []
    n_comp = range(1,64)
    for i in n_comp:
        pca = PCA(n_components=i)
        pca.fit(train_images2)

        tX_train = pca.transform(train_images2)
        tX_test = pca.transform(test_images2)

        model = GaussianNB().fit(tX_train, train_labels)
        test_score = model.score(tX_test, test_labels)
        accuracy.append(test_score)

py.plot(n_comp, accuracy)
    py.xlabel('number of PCA components')
    py.ylabel('digit recognition accuracy')
```

Out[101]: Text(0, 0.5, 'digit recognition accuracy')



```
In [103]: import matplotlib.pyplot as plt
          import numpy as np
          # Import datasets, classifiers and performance metrics
          from sklearn import datasets, metrics
          from sklearn.model selection import GridSearchCV, train test split, cross val score
          # import warnings filter
          from warnings import simplefilter
          # ignore all future warnings
          simplefilter(action='ignore', category=FutureWarning)
          final = [[""]*4 for i in range(3)]
          fashion_mnist = keras.datasets.fashion_mnist
           (train images, train labels), (test images, test labels) = fashion mnist.load da
          n samples = len(train images)
          expected = train_labels
          data = train images.reshape((n samples, -1))
          X_train, X_test, y_train, y_test = train_test_split(data, expected, test_size=0.4
```

```
In [107]: from sklearn.neighbors import KNeighborsClassifier
    parameters={'n_neighbors':[2,5]}
    knn = KNeighborsClassifier()
    scores = cross_val_score(knn, X_train, y_train, cv=5)
    default_train=np.mean(scores)

    clf=GridSearchCV(knn, parameters, cv=5,scoring="accuracy",iid=False)
    clf.fit(X_train,y_train)
    best_train=clf.best_score_
    nn=clf.best_estimator_.n_neighbors

    y_pred = clf.predict(X_test) # runs the model with the best score
    best_test=metrics.accuracy_score(y_test,y_pred)

    final[0][0]=str(default_train)
    final[0][1]=str(best_train)
    final[0][2]="n_neighbors: "+str(nn)
    final[0][3]=str(best_test)
```

```
In [108]:
          from sklearn import tree
          DT=tree.DecisionTreeClassifier()
          scores=cross_val_score(DT, X_train, y_train, cv=5)
          default train=np.mean(scores)
          parameters={'min_samples_split':list(range(2,10))}
          clf=GridSearchCV(DT, parameters, cv=5,scoring="accuracy",iid=False)
          clf.fit(X train,y train)
          best train=clf.best score
          min_samples=clf.best_estimator_.min_samples_split
          y_pred = clf.predict(X_test) # runs the model with the best score
          best_test=metrics.accuracy_score(y_test,y_pred)
          final[1][0]=str(default train)
          final[1][1]=str(best_train)
          final[1][2]="min samples: "+str(min samples)
          final[1][3]=str(best_test)
```

```
In [109]: from sklearn.linear_model import LogisticRegression
    regression=LogisticRegression(penalty='l1', solver='saga', tol=0.01)
    scores=cross_val_score(regression, X_train, y_train, cv=5)
    default_train=np.mean(scores)

parameters={'C':[10**(i-4) for i in range(9)]}
    clf=GridSearchCV(regression, parameters, cv=5,scoring="accuracy",iid=False)
    clf.fit(X_train,y_train)
    best_train=clf.best_score_
    C=clf.best_estimator_.C

y_pred = clf.predict(X_test) # runs the model with the best score
    best_test=metrics.accuracy_score(y_test,y_pred)

final[2][0]=str(default_train)
    final[2][1]=str(best_train)
    final[2][2]="C: "+str(C)
    final[2][3]=str(best_test)
```

```
In [110]: left=["K-nearest neighbors: ","Decision Trees:
                                                              ","Logistic Regression: "]
                                       "," Default Accuracy", "Tuned Valid. Acc.",
          print("Model
           " Hyperparameters "," Test Set Acc. ",sep="|",end="|\n")
          for i in range(50):
          print("_",end="_")
print("")
          for i in range(3):
           print(left[i],end="|")
           for j in range(4):
             print('{:>17.15}'.format(final[i][j]),end="|")
           print("")
          Model
                               | Default Accuracy|Tuned Valid. Acc. | Hyperparameters |
          est Set Acc. |
          K-nearest neighbors:
                                  0.8450288399279
                                                                        n neighbors: 5
                                                    0.8450288399279
          0.8509583333333
          Decision Trees:
                                  0.7829170194727
                                                    0.7844444731603
                                                                       min samples: 9
          0.7865
          Logistic Regression: | 0.8455008694235 | 0.8456952599081 |
                                                                                 C: 1
          0.850375
```