## nn\_analysis-Copy1

## October 11, 2020

```
[1]: import six
     import sys
     sys.modules['sklearn.externals.six'] = six
     import mlrose
     from sklearn import preprocessing
     from sklearn import metrics
     from sklearn.model_selection import train_test_split
     from sklearn.neural_network import MLPClassifier
     import time
     import random
     import numpy as np
     import pandas as pd
     import matplotlib.pyplot as plt
     %matplotlib inline
[2]: def import_data():
         df_ds1 = pd.read_csv("./winequality-red.csv", sep=";").append(pd.read_csv(".
      →/winequality-white.csv", sep=";"))
         missing_values = ['?']
         df_ds2 = pd.read_csv("./breast-cancer-wisconsin.csv", sep=",",na_values =_
     →missing_values)
         df_ds2.fillna(method='ffill',inplace=True)
         X_ds1 = np.array(df_ds1.values[:,1:-1])
         y_ds1 = np.array(df_ds1.values[:,-1])
         y_ds1 = (y_ds1<7).astype(int)
         X_ds2 = np.array(df_ds2.values[:,1:-1])
         y_ds2 = np.array(df_ds2.values[:,-1])
         return X_ds1, y_ds1, X_ds2, y_ds2
[3]: def train_test(X, y):
         X_train, X_test, y_train, y_test = train_test_split(np.array(X),np.
      →array(y), test_size=0.20)
         sc = preprocessing.StandardScaler()
         X_train = sc.fit_transform(X_train)
         X_test = sc.transform(X_test)
         return X_train, X_test, y_train, y_test
```

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[4]: def nn_learner(X_train, X_test, y_train, y_test):
         \#clf\_nn = MLPClassifier(hidden\_layer\_sizes=(10, 10, 10), random\_state=18, 
      \rightarrow max_iter=1000)
         __clf_nn = MLPClassifier(hidden_layer_sizes=(20, 20,10),random_state=18)
         print(__clf_nn)
         __clf_nn.fit(X_train, y_train)
         y_pred = __clf_nn.predict(X_test)
         nn_accuracy = metrics.accuracy_score(y_test, y_pred)
         print('Accuracy of neural network without hyperparameter tuning is %.2f%%'\_
      \rightarrow% (nn_accuracy * 100))
         return __clf_nn
[5]: def___
      → __nn_analysis(algorithm, __hidden_layer, __learning_rate, max_iters, max_attempts, __
      →X_train, X_test, y_train, y_test,\
                         pop_size = 200, mutation_prob = 0.2,restarts_
      ⇒=0, schedule=mlrose. GeomDecay(), random_state=None):
         if algorithm == 'simulated_annealing':
             nn_model1 = mlrose.NeuralNetwork(hidden_nodes = __hidden_layer,__
      →activation = 'relu', algorithm = algorithm,
                                     max_iters = max_iters, bias = True,
      →is_classifier = True, learning_rate = __learning_rate,
      -early_stopping=True,max_attempts=max_attempts,schedule=schedule,random_state=random_state)
         elif algorithm == 'genetic_alg':
             nn_model1 = mlrose.NeuralNetwork(hidden_nodes = __hidden_layer,__
      →activation = 'relu', algorithm = algorithm,
                                      max_iters = max_iters, bias = True,__
      →is_classifier = True, learning_rate = __learning_rate,
                                      early_stopping=True,max_attempts=max_attempts,\
                                      pop_size = pop_size, mutation_prob = __
      →mutation_prob,random_state=random_state)
         elif algorithm == 'random_hill_climb':
             nn model1 = mlrose.NeuralNetwork(hidden_nodes = __hidden_layer,__
      →activation = 'relu', algorithm = algorithm,
                                     max_iters = max_iters, bias = True,
      →is_classifier = True, learning_rate = __learning_rate,
      →early_stopping=True,max_attempts=max_attempts,restarts_
      ⇒=restarts,random_state=random_state)
         elif algorithm == 'gradient_descent':
             nn_model1 = mlrose.NeuralNetwork(hidden_nodes = __hidden_layer,__
      →activation = 'relu', algorithm = algorithm,
                                      max_iters = max_iters, bias = True,
      →is_classifier = True, learning_rate = __learning_rate,
```

```
early_stopping=True,_
      →max_attempts=max_attempts,restarts =restarts,random_state=random_state)
         start_time_train = time.time()
         nn model1.fit(X train, y train)
         end_time_train = time.time() - start_time_train
         # Predict labels for train set and assess accuracy
         y_train_pred = nn_model1.predict(X_train)
         y_train_accuracy = metrics.accuracy_score(y_train, y_train_pred)
         print('Training accuracy: ', y_train_accuracy)
         # Predict labels for test set and assess accuracy
         start_time_pred = time.time()
         y_test_pred = nn_model1.predict(X_test)
         end_time_pred = time.time() - start_time_pred
         y_test_accuracy = metrics.accuracy_score(y_test, y_test_pred)
         print('Test accuracy: ', y_test_accuracy)
         return y_train_accuracy, y_test_accuracy, end_time_train, end_time_pred
[6]: def train_nn_optimization(_hidden_layer,__learning_rate,max_attempts, X_train,_u
     →X_test,y_train, y_test,\
                                 __range=range(1000,10000,1000), pop_size=200,__
     mutation_prob=0.2,restarts =0,schedule=mlrose.GeomDecay(),random_state=None):
         __nn_metrics = {}
         for algorithm in ['random_hill_climb', 'simulated_annealing', _
             __train_accuracy_values, __test_accuracy_values,__train_time_values,_
      →__pred_time_values = [],[],[],[]
             for i in __range:
                 __train_accuracy, __test_accuracy,__train_time, __pred_time =_
     → __nn_analysis(algorithm, __hidden_layer, __learning_rate, i, \
                 max_attempts, X_train, X_test, \
                 y_train, y_test,pop_size, \
                 mutation_prob,restarts,schedule,random_state)
                 __train_accuracy_values.append(__train_accuracy)
                 __test_accuracy_values.append(__test_accuracy)
                 __train_time_values.append(__train_time)
```

\_\_pred\_time\_values.append(\_\_pred\_time)

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__nn_metrics[algorithm] = {'train_accuracy' : __train_accuracy_values,_\u00cd

\u20f3'test_accuracy' : __test_accuracy_values, 'train_time' :_\u00cd

\u20f3_train_time_values, 'pred_time' : __pred_time_values}
\u20f3
return __nn_metrics
```

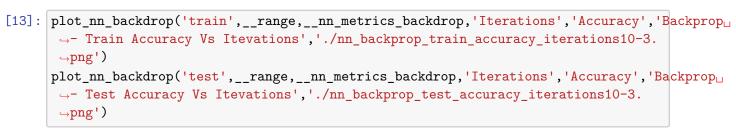
```
[7]: def___

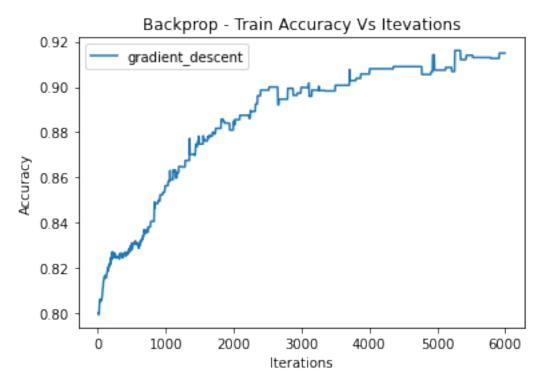
¬plot_nn_optimization(plot_type,y_axis,results,x_label,y_label,title,filename):
        for algorithm in ['random_hill_climb', 'simulated_annealing',_
     if plot_type == 'train' :
                plt.plot(y_axis,np.
     →array(results[algorithm]['train_accuracy']),label=algorithm)
             elif plot type == 'test' :
                plt.plot(y_axis,np.
     →array(results[algorithm]['test_accuracy']),label=algorithm)
             elif plot_type == 'time' :
                plt.plot(y_axis,np.
     →array(results[algorithm]['train_time']),label=algorithm)
             elif plot_type == 'pred' :
                plt.plot(y_axis,np.
     →array(results[algorithm]['pred_time']),label=algorithm)
        plt.xlabel(x label)
        plt.ylabel(y_label)
        plt.title(title)
        plt.legend(loc='best')
        plt.savefig(filename)
        plt.show()
        plt.clf()
```

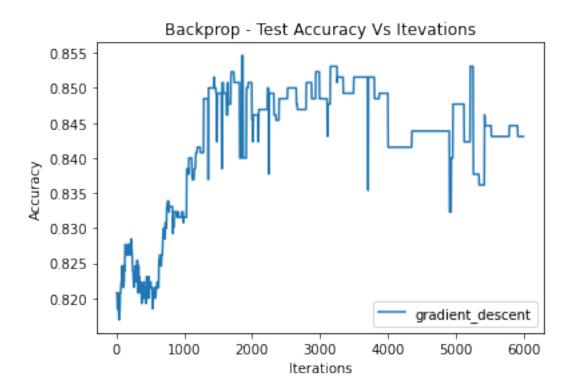
```
mutation_prob,restarts,schedule,random_state)
              __train_accuracy_values.append(__train_accuracy)
              __test_accuracy_values.append(__test_accuracy)
              __train_time_values.append(__train_time)
              __pred_time_values.append(__pred_time)
          __nn_metrics[algorithm] = {'train_accuracy' : __train_accuracy_values,_
       \hookrightarrow 'test_accuracy' : __test_accuracy_values, 'train_time' :_
       →__train_time_values, 'pred_time' : __pred_time_values}
          return __nn_metrics
 [9]: def plot_nn_backdrop(plot_type,y_axis,results,x_label,y_label,title,filename):
          algorithm = 'gradient_descent'
          if plot_type == 'train' :
              plt.plot(y_axis,np.
       →array(results[algorithm]['train_accuracy']),label=algorithm)
          elif plot_type == 'test' :
              plt.plot(y_axis,np.
       →array(results[algorithm]['test_accuracy']),label=algorithm)
          plt.xlabel(x label)
          plt.ylabel(y_label)
          plt.title(title)
          plt.legend(loc='best')
          plt.savefig(filename)
          plt.show()
          plt.clf()
[10]: def plot_nn_backdrop_both(y_axis,results,x_label,y_label,title,filename):
          algorithm = 'gradient_descent'
          plt.plot(y_axis,np.
       →array(results[algorithm]['train_accuracy']),label=algorithm + ' - train')
          plt.plot(y_axis,np.
       array(results[algorithm]['test_accuracy']),label=algorithm + ' - test')
          plt.xlabel(x_label)
          plt.ylabel(y_label)
          plt.title(title)
          plt.legend(loc='best')
          plt.savefig(filename)
          plt.show()
          plt.clf()
[11]: X_ds1, y_ds1, X_ds2, y_ds2 = import_data()
      X_train, X_test, y_train, y_test = train_test(X_ds1, y_ds1)
      __clf_nn = nn_learner(X_train, X_test, y_train, y_test)
```

MLPClassifier(hidden\_layer\_sizes=(20, 20, 10), random\_state=18)
Accuracy of neural network without hyperparameter tuning is 83.15%

C:\Dev\Anaconda3\envs\VitrualEnv\lib\sitepackages\sklearn\neural\_network\\_multilayer\_perceptron.py:582:
ConvergenceWarning: Stochastic Optimizer: Maximum iterations (200) reached and the optimization hasn't converged yet.
 warnings.warn(





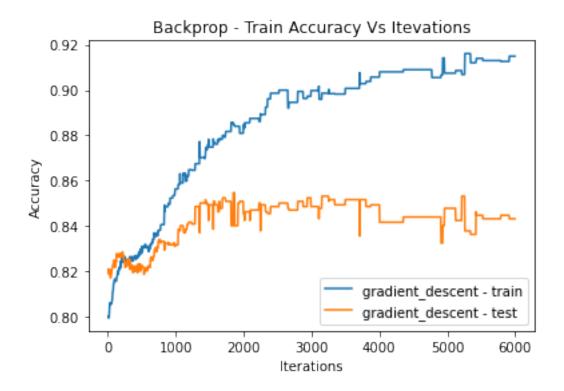


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[14]: plot_nn_backdrop_both(__range,__nn_metrics_backdrop,'Iterations','Accuracy','Backprop_

→ Train Accuracy Vs Itevations','./

→nn_backprop_train_accuracy_iterations10_both-3.png')
```



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```
[]: max_attempts = 100
    __range=range(1000,100001,10000)
    __schedule = mlrose.GeomDecay()
    __hidden_layer = list(__clf_nn.hidden_layer_sizes)
#__hidden_layer = [5]
#__learning_rate = __clf_nn.learning_rate_init
    __learning_rate = 0.1

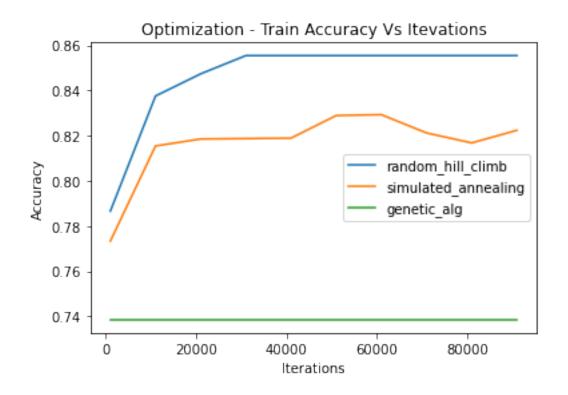
__nn_metrics_optimization =_u
    __train_nn_optimization(_hidden_layer,__learning_rate,max_attempts, X_train,_u
    __\text{__itest_,y_train, y_test_,__range, 200, 0.2,0,mlrose.GeomDecay(),18)}
```

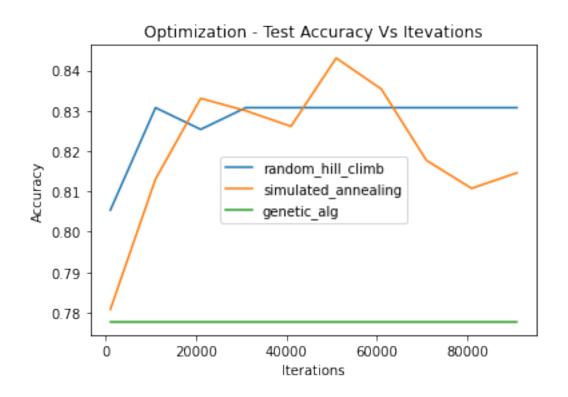
```
plot_nn_optimization('train',__range,__nn_metrics_optimization,'Iterations','Accuracy','Optimization_train_accuracy_iterations10-3.png')

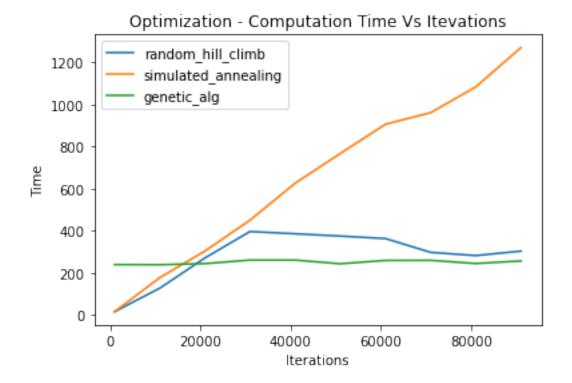
plot_nn_optimization('test',__range,__nn_metrics_optimization,'Iterations','Accuracy','Optimization_optimization_test_accuracy_iterations10-3.png')

plot_nn_optimization_test_accuracy_iterations10-3.png')

plot_nn_optimization('time',__range,__nn_metrics_optimization,'Iterations','Time','Optimization_optimization_time','Optimization_optimization_time_accuracy_iterations10-3.png')
```







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