

# **Computational Intelligence & Adversarial Machine Learning:**

Assignment #3: Genetic & Evolutionary Feature Selection (GEFeS)

# Assignment #3

(Due 11/1/2018)

## Assignment #3

- Given the baselines for an Linear Support Vector Machine (SVM), a Radial-Based Support Vector Machine (RBSVM) and a Backpropagation Neural Network select one of these three to improve upon the baseline accuracy using the following GEFes Methods (to evolve a population of Feature Masks):
  - A Steady-State GA
  - An Elitist GA
  - An Estimation of Distribution Algorithm
- Your three GEFes methods should use the following parameters:
  - Population Size = 25
  - Crossover = Uniform
  - Mutation Rate = 0.05
  - Number of Evaluations = 5000
  - Evaluation Function = Accuracy
- You should run your GEFes methods 10 times and record the average and best results.

# Assignment #3

(Due 11/1/2017)

## Assignment #3 (Grading)

- [30pts] Developing the three GEFes methods
- [15pts] For the a table of the performances of the three methods recording the average and best performance of each method
- [15pts] For a plot of the Accuracy Curves with respect to the best performances of your three methods
- [10pts] For any innovation for improving GEFes methods (one innovation per GEFes method)
- [30pts] Develop the paper.

# Assignment #3

(Due 11/1/2017)

## Assignment #3 (cont.)

- Write a paper using IEEE or AAAI format documenting your work:
  - I. Title
  - II. Authors
  - III. Abstract
  - IV. Introduction
  - V. Methodology
  - VI. Experiment
  - VII. Results
  - VIII. Breakdown of the Work
  - IX. References

# Assignment #3

(Due 11/1/2018)

```
import Data_Utils
from sklearn.model_selection import StratifiedKFold
from sklearn.preprocessing import StandardScaler, normalize
from sklearn import svm
from sklearn.neural_network import MLPClassifier
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.model_selection import StratifiedKFold
from sklearn import preprocessing
import numpy as np
from sklearn.model_selection import cross_val_score
```

```
CU_X, Y = Data_Utils.Get_Casis_CUDataset()
```

```
rbfsvm = svm.SVC()
lsvm = svm.LinearSVC()
mlp = MLPClassifier(max_iter=2000)
```

```
skf = StratifiedKFold(n_splits=4, shuffle=True, random_state=0)
fold_accuracy = []
```

```
scaler = StandardScaler()
tfidf = TfidfTransformer(norm=None)
dense = Data_Utils.DenseTransformer()
```

```
for train, test in skf.split(CU_X, Y):
    #train split
    CU_train_data = CU_X[train]
    train_labels = Y[train]
```

```
    #test split
    CU_eval_data = CU_X[test]
    eval_labels = Y[test]
```

```
    # tf-idf
    tfidf.fit(CU_train_data)
    CU_train_data = dense.transform(tfidf.transform(CU_train_data))
    CU_eval_data = dense.transform(tfidf.transform(CU_eval_data))
```

```
    # standardization
    scaler.fit(CU_train_data)
    CU_train_data = scaler.transform(CU_train_data)
    CU_eval_data = scaler.transform(CU_eval_data)
```

```
    # normalization
    CU_train_data = normalize(CU_train_data)
    CU_eval_data = normalize(CU_eval_data)
```

```
    train_data = CU_train_data
    eval_data = CU_eval_data
```

```
    # evaluation
    rbfsvm.fit(train_data, train_labels)
    lsvm.fit(train_data, train_labels)
    mlp.fit(train_data, train_labels)
```

```
    rbfsvm_acc = rbfsvm.score(eval_data, eval_labels)
    lsvm_acc = lsvm.score(eval_data, eval_labels)
    mlp_acc = mlp.score(eval_data, eval_labels)
```

```
    fold_accuracy.append((lsvm_acc, rbfsvm_acc, mlp_acc))
```

```
print(np.mean(fold_accuracy, axis = 0))
```

Baseline Accuracies = [0.73 0.63 0.67]



**Have a Great Day!!!**