

Assignment-2

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Support vector Machine Assignment

Binary classification of first two classes with first 10 features: `svm_image_classifier.py`

```
C_range = 10. ** np.arange(-1, 3)
gamma_range = 10. ** np.arange(-8, -3)
degree_range = np.arange(1, 4)
```

with linear kernel:

train accuracy = 0.99, test accuracy = 0.983333333333, same for all C values

With poly: used grid search for optimal values.

optimal values: C=0.001, gamma=0.001, degree=1, train score =0.991666666667, test score=1.0

after compiling `svm_image_classifier.py` code you will see output as below,

accuracy: 0.98958, params: {'kernel': 'poly', 'C': 1.0, 'degree': 3, 'max_iter': 100, 'coef0': -16, 'gamma': 9.999999999999995e-08}

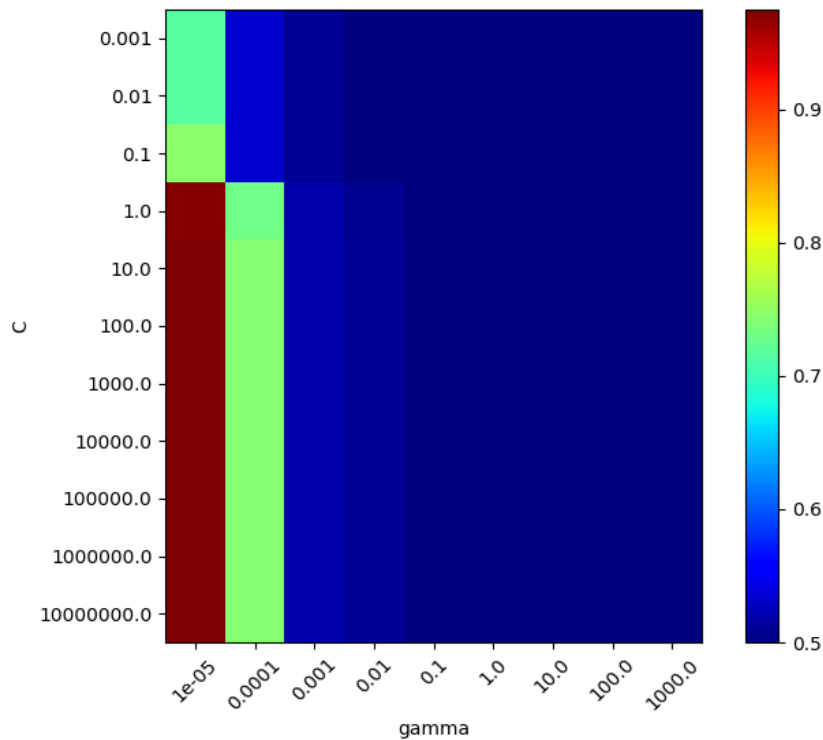
accuracy: 0.98125, params: {'kernel': 'poly', 'C': 1.0, 'degree': 3, 'max_iter': 100, 'coef0': -16, 'gamma': 9.999999999999995e-07}

So we can see that as gamma is increasing validation error increasing which means overfitting

With rbf: used grid search to obtain optimal values

C=10.0, gamma=0.00001, train score =1, test score= 0.983333333333.

C vs gamma



Binary classification of first two classes with all features: `svm_all_features_binary.py`

```
C_range = 10. ** np.arange(-3, 8)
gamma_range = 10. ** np.arange(-5, 4)
degree_range = np.arange(1,10)
```

with linear kernel:

train accuracy = 0.75, test accuracy = 0.746, same for all C values

With poly: used grid search for optimal values. Optimal values are C=0.001, gamma=0.001,

degree=3, train accuracy= 0.9979166666667, test accuracy = 0.9916666666667

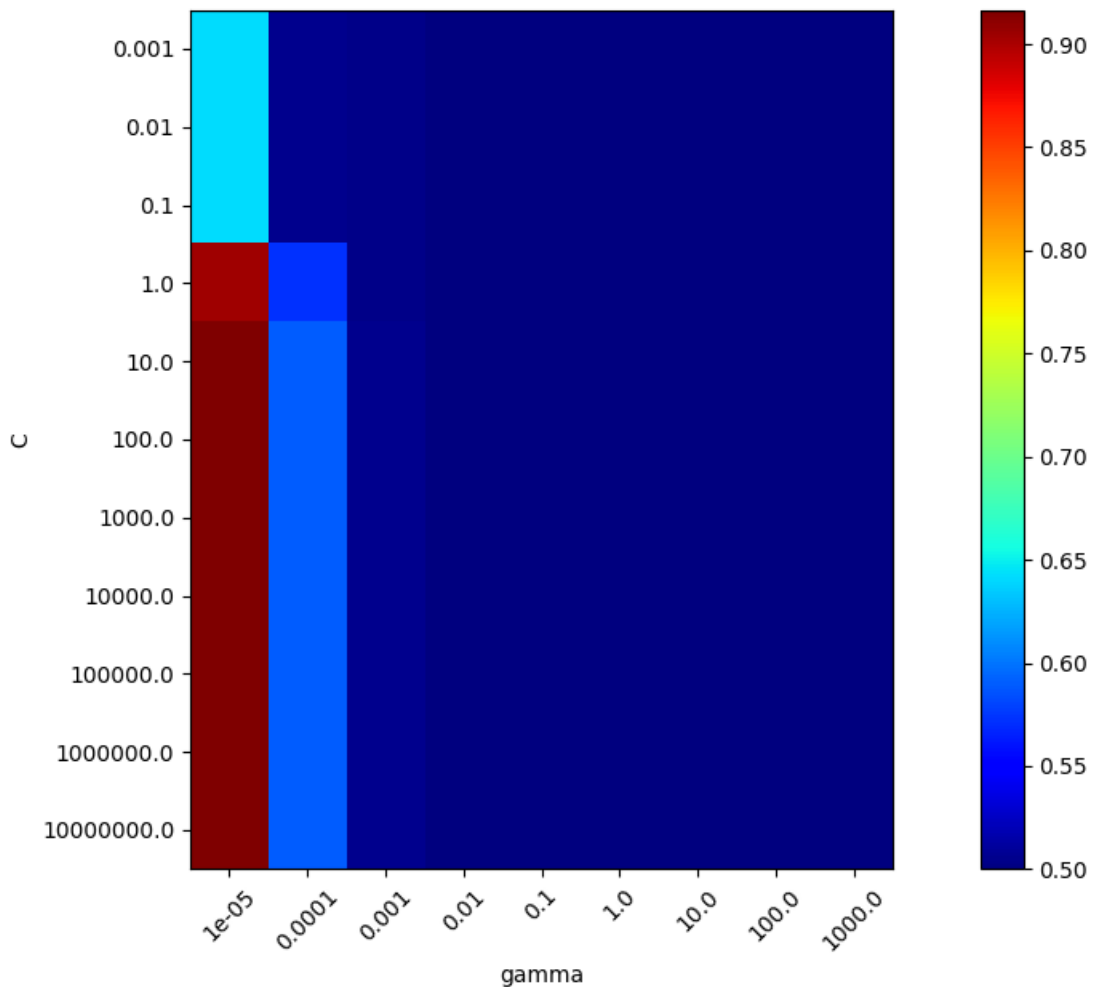
mean: 0.97292, params: {'kernel': 'poly', 'C': 10000000.0, 'degree': 2, 'max_iter': 10, 'coef0': -7, 'gamma': 0.001}

mean: 0.98750, params: {'kernel': 'poly', 'C': 10000000.0, 'degree': 2, 'max_iter': 10, 'coef0': -7, 'gamma': 0.0001}

So we can see that as gamma is increasing validation error increasing which means underfitting

With rbf: C=10.0, gamma=0.00001, train accuracy= 0.9166666666667, test accuracy=0.925

C vs gamma



Multi-Classification with 10 features: `svm_multi_10features.py`

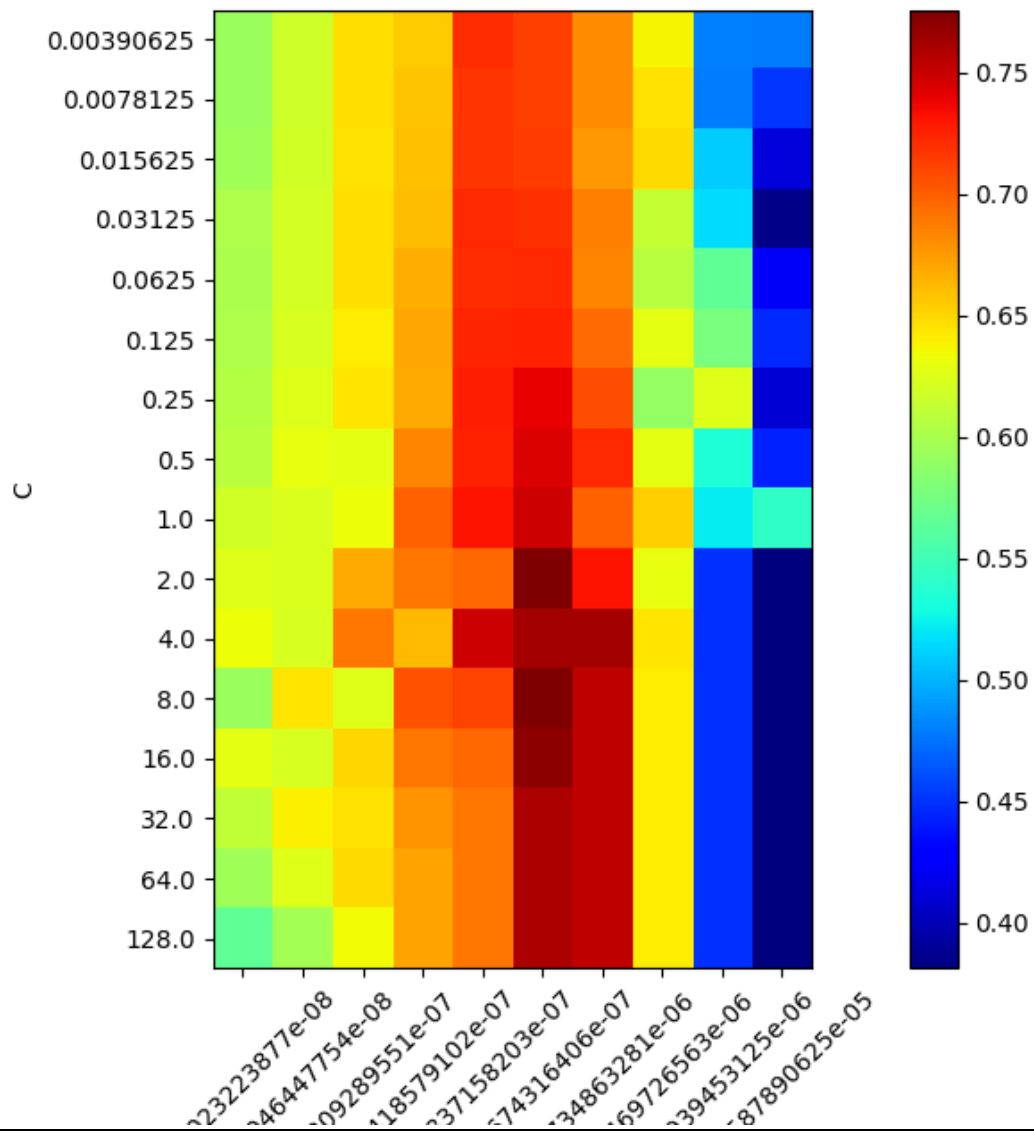
```
C_range = 2. ** np.arange(-8, 8)
gamma_range = 2. ** np.arange(-25, -15)
degree_range = np.arange(1, 20)
```

With linear: train accuracy = 0.41, test accuracy = 0.39, almost same for all C values

with poly: C=0.03125, gamma=9.5367e-07, training accuracy=0.67, testing accuracy = 0.72, degree=3

with rbf: train accuracy = 0.770833333333, test accuracy = 0.74, C=8.0, gamma=9.5367431640625e-07

C vs gamma

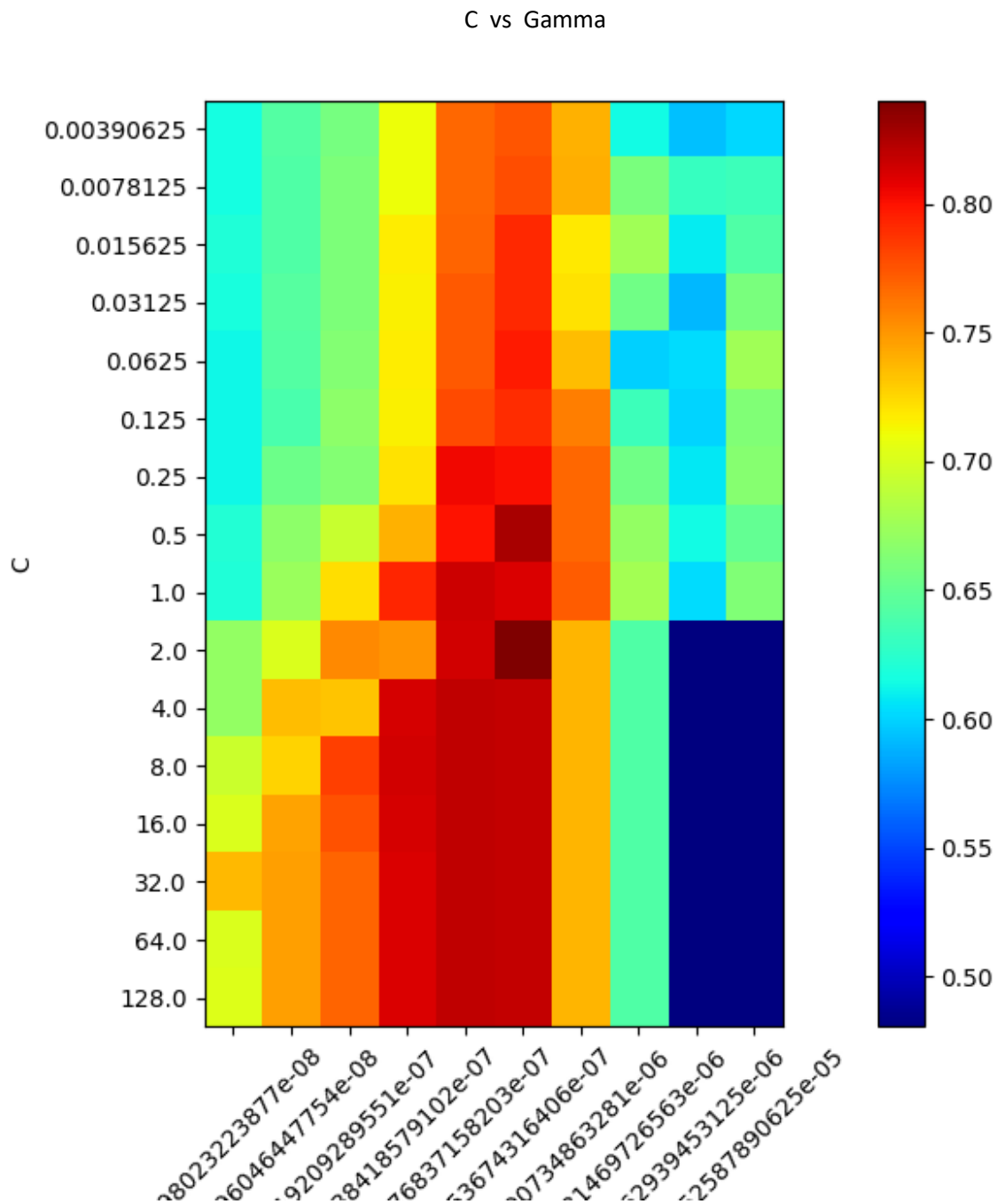


Multi-Classification with all features: svm_multiu.py

```
C_range = 2. ** np.arange(-8, 8)
gamma_range = 2. ** np.arange(-25, -15)
degree_range = np.arange(1, 20)
```

With linear: Training Accuracy = 0.63, Testing accuracy = 0.57, almost no change for all values of C
with poly: C=0.015625, gamma=9.5367e-07, training accuracy=0.77, testing accuracy = 0.72, degree=5

With rbf: C= 2.0, gamma= 9.5367431640625e-07, Training Accuracy = 0.844166666667,
Testing accuracy = 0.821666666667.



Part -2: part2.py

with linear kernel: Training Accuracy = 0.52340416213, Testing accuracy = 0.51215484446.

So linear kernel is underfitting or unable to learn the model accurately no matter what are the values of C . This is because of large dimensions of data

The below are the examined C , gamma and degree (for poly kernel) ranges .

```
C_range = 2. ** np.arange(-5, 15)
gamma_range = 2. ** np.arange(-15, 3)
degree_range = np.arange(1,20)
```

With poly: $C=0.03125$, $\gamma=2.4414062e-05$, training accuracy=0.805, testing accuracy = 0.79, degree=8

With rbf:

$C=0.0625$, $\gamma=6.103515625e-05$, training accuracy=0.925, testing accuracy = 0.915

C vs gamma

