

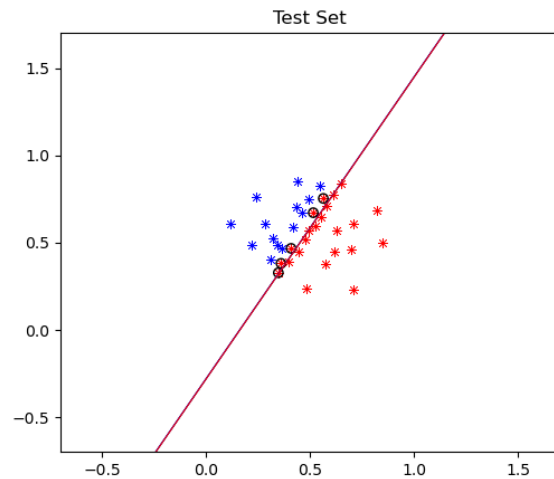
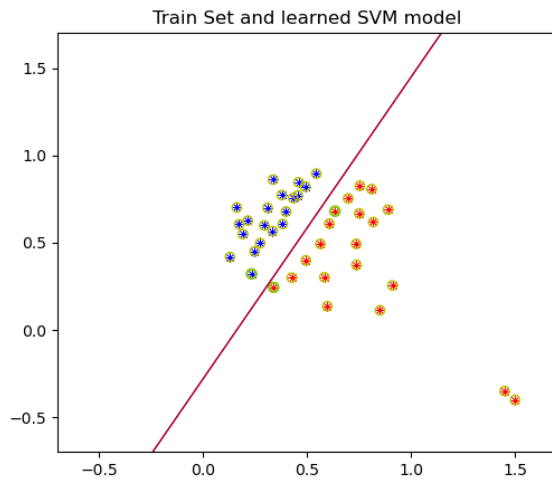
## Question 1

(a)

Number of SV: 3

Accuracy on train data with  $C = 1000$ : 1.0

Accuracy on test data with  $C = 1000$ : 0.8648648648648649



### Adjustments to `apply_a.py`

`python apply_a.py` threw a lot of errors, which were mainly due to not enforcing a datatype (i.e. matrix vs nparray) and not taking the dimensions requested in `simlin.py` into account.

For this reason, we have modified `apply_a.py`.

(b+c)

**1-vs-3**

Number of SV: 26

Width of margin: 0.000527852066940575

Train Error 1-vs-3: 7

Test Error 1-vs-3: 0.013953488372093023

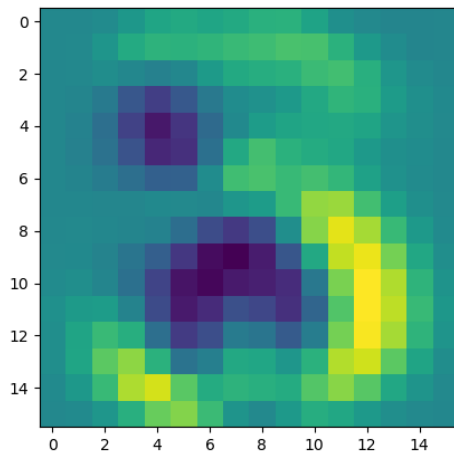
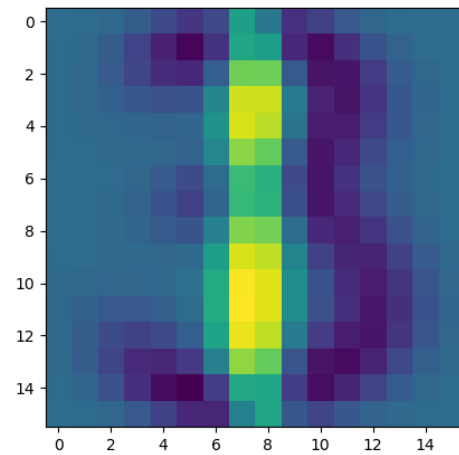
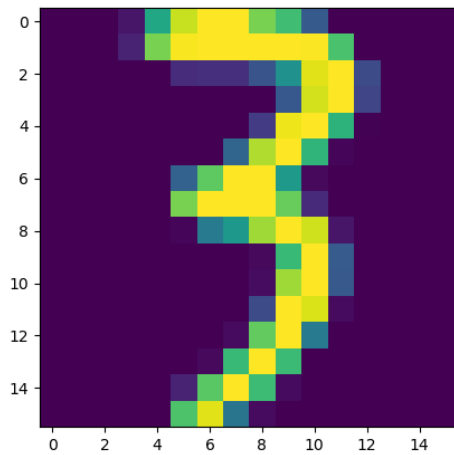
**3-vs-8**

Number of SV: 89

Width of margin: 0.00010191438464165348

Train Error 3-vs-8: 66

Test Error 3-vs-8: 0.09939759036144578



### Adjustments to `apply_bc.py`

As `python apply_a.py`, `python apply_bc.py` threw a lot of errors, which were mainly due to not enforcing a datatype (i.e. matrix vs nparray) and not taking the dimensions requested in `simlin.py` into account.

For this reason, we have modified `apply_bc.py`.

(d)

Accuracy of kernel SVM with  $C=1000$  and  $\text{norm}=5$ : 0.40540540540540543

This shows, that the accuracy of the kernel SVM is much lower than the accuracy of the linear SVM.

### Adjustments to `apply_d.py` and `svmkern.py`

There was a mismatch between the number of variables unpacked from the function `svmkern` in `apply_d.py` and the number of variables to be returned by this function, defined by its description. We have extended the function to return the same values as `svmlin`.

As with the previous adjustments that had to be made, we again had to enforce datatypes and dimensions.