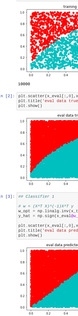
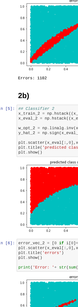
2a)

In [1]:
```
import numpy as np
from scipy.io import loadmat
import matplotlib.pyplot as plt

in_data = loadmat('classifier_data.mat')
#print([key for key in in_data]) # <- use this line to see the keys in the dictionary data s
#uncomment

x_train = in_data['x_train']
x_eval = in_data['x_eval']
y_train = in_data['y_train']
y_eval = in_data['y_eval']

n_eval = np.size(y_eval)
n_train = np.size(y_train)

plt.scatter(x_train[:,0],x_train[:,1], color='c' if i==-1 else 'r' for i in y_train[:,0]])
plt.title('training data')
plt.show()
print(n_eval)
```




```
10000
```

In [2]:
```
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==-1 else 'r' for i in y_eval[:,0]])
plt.title('eval data true class')
plt.show()
```



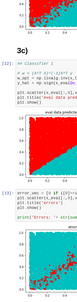
In [3]:
```
## Classifier 1

# w = (XT*X)^(-1)XT*y
w_opt = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_eval@w_opt)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat[:,0]])
plt.title('eval data predicted class (y_hat)')
plt.show()
```

In [4]:
```
error_vec = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec])
plt.title('errors')
plt.show()

print('Errors: ' + str(sum(error_vec)))
```

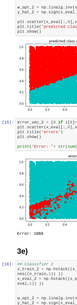

```
Errors: 1102
```

2b)

In [5]:
```
## Classifier 2
x_train_2 = np.hstack((x_train**2, x_train, np.ones((n_train,1)) ))
x_eval_2 = np.hstack((x_eval**2, x_eval, np.ones((n_eval,1)) ))

w_opt_2 = np.linalg.inv(x_train_2.transpose()@x_train_2.transpose())@y_train
y_hat_2 = np.sign(x_eval_2@w_opt_2)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat_2[:,0]])
plt.title('predicted class on eval data')
plt.show()
```



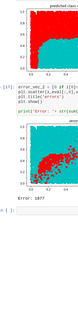
In [6]:
```
error_vec_2 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat_2, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec_2])
plt.title('errors')
plt.show()

print('Error: ' + str(sum(error_vec_2)))
```




```
Error: 542
```

2c)

In [7]:
```
## create new, correctly labeled points
n_new = 1000 #number of new datapoints
x_train_new = np.hstack((np.zeros((n_new,1)), 3*np.ones((n_new,1))))
y_train_new = np.ones((n_new,1))

## add these to the training data
x_train_outlier = np.vstack((x_train,x_train_new))
y_train_outlier = np.vstack((y_train,y_train_new))
plt.scatter(x_train_outlier[:,0],x_train_outlier[:,1], color='c' if i==-1 else 'r' for i in
y_train_outlier[:,0]])
plt.title('new training data')
plt.show()
```



In [8]:
```
#train with new data
w_opt_outlier = np.linalg.inv(x_train_outlier.transpose()@x_train_outlier)@x_train_outlier.t
ranspose()@y_train_outlier
y_hat_outlier = np.sign(x_eval@w_opt_outlier)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat_outlier[:,0
]])
plt.title('predicted class on eval data')
plt.show()
```


In [9]:
```
error_vec = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat_outlier, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec])
plt.title('errors')
plt.show()

print('Errors: ' + str(sum(error_vec)))
```



```
Errors: 2134
```

3a)

In [10]:
```
import numpy as np
from scipy.io import loadmat
import matplotlib.pyplot as plt

in_data = loadmat('overfitting_data.mat')
#print([key for key in in_data]) # <- use this line to see the keys in the dictionary data s
#uncomment

x_train = in_data['x_train']
x_eval = in_data['x_eval']
y_train = in_data['y_train']
y_eval = in_data['y_eval']

n_eval = np.size(y_eval)
n_train = np.size(y_train)

plt.scatter(x_train[:,0],x_train[:,1], color='c' if i==-1 else 'r' for i in y_train[:,0]])
plt.title('training data')
plt.show()
```


3b)

In [11]:
```
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==-1 else 'r' for i in y_eval[:,0]])
plt.title('eval data true class')
plt.show()
```


3c)

In [12]:
```
## Classifier 1

# w = (XT*X)^(-1)XT*y
w_opt = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_eval@w_opt)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat[:,0]])
plt.title('eval data predicted class (y_hat)')
plt.show()
```


In [13]:
```
error_vec = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec])
plt.title('errors')
plt.show()

print('Errors: ' + str(sum(error_vec)))
```


```
Errors: 759
```

3d)

In [14]:
```
## Classifier 2
x_train_2 = np.hstack((x_train**2, x_train, np.ones((n_train,1)) ))
x_eval_2 = np.hstack((x_eval**2, x_eval, np.ones((n_eval,1)) ))

w_opt_2 = np.linalg.inv(x_train_2.transpose()@x_train_2@x_train_2.transpose())@x_train
y_hat_2 = np.sign(x_eval_2@w_opt_2)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat_2[:,0]])
plt.title('predicted class on eval data')
plt.show()
```

In [15]:
```
error_vec_2 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat_2, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec_2])
plt.title('errors')
plt.show()

print('Error: ' + str(sum(error_vec_2)))
```



```
Error: 1006
```

3e)

In [16]:
```
## Classifier 3
x_train_3 = np.hstack((x_train**3,x_train**2,x_train**2,x_train**3,x_train**2, x_train, np.o
nes((n_train,1)) ))
x_eval_3 = np.hstack((x_eval**3,x_eval**2,x_eval**2,x_eval**3,x_eval**2, x_eval,  np.ones((n_
eval,1)) ))

w_opt_3 = np.linalg.inv(x_train_3.transpose()@x_train_3@x_train_3.transpose())@y_train
y_hat_3 = np.sign(x_eval_3@w_opt_3)

plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==1 else 'r' for i in y_hat_3[:,0]])
plt.title('predicted class on eval data')
plt.show()
```


In [17]:
```
error_vec_3 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat_3, y_eval))]
plt.scatter(x_eval[:,0],x_eval[:,1], color='c' if i==0 else 'r' for i in error_vec_3])
plt.title('errors')
plt.show()

print('Error: ' + str(sum(error_vec_3)))
```



```
Error: 1077
```

In [ ]:
```

```