# **Question 4**

```
In [49]:
```

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
import math
```

#### In [50]:

```
from cvxopt import matrix, solvers #For Quadratic Programming
```

#### In [51]:

```
from sympy import *
```

# **Reading Iris Dataset**

#### In [52]:

```
iris_ds= pd.read_csv("Iris_dataset.csv",index_col=False)
print(iris_ds.head())
```

	sepal.length	sepal.width	petal.length	petal.width	variety
0	5.1	3.5	1.4	0.2	Setosa
1	4.9	3.0	1.4	0.2	Setosa
2	4.7	3.2	1.3	0.2	Setosa
3	4.6	3.1	1.5	0.2	Setosa
4	5.0	3.6	1.4	0.2	Setosa

# Taking the two features only

## In [53]:

```
iris_data = iris_ds.drop(['sepal.length','petal.width'], axis=1)
print(iris_data)
```

```
sepal.width petal.length
                                   variety
0
             3.5
                            1.4
                                    Setosa
1
             3.0
                            1.4
                                    Setosa
2
             3.2
                            1.3
                                    Setosa
3
             3.1
                            1.5
                                    Setosa
4
             3.6
                            1.4
                                    Setosa
                            . . .
145
             3.0
                            5.2 Virginica
             2.5
                            5.0 Virginica
146
147
             3.0
                            5.2 Virginica
148
             3.4
                            5.4 Virginica
149
             3.0
                            5.1 Virginica
```

[150 rows x 3 columns]

Dividing the Data into 3 sets (Setosa vs Versicolor, Setosa vs Virginica, Versicolor vs Virginica Since Single Perceptron and SVM can handle two classes

#### In [54]:

```
iris_data_12 = iris_data[iris_data['variety']!='Virginica']
print(iris_data_12)
```

	sepal.width	petal.length	variety
0	3.5	1.4	Setosa
1	3.0	1.4	Setosa
2	3.2	1.3	Setosa
3	3.1	1.5	Setosa
4	3.6	1.4	Setosa
	• • •	• • •	
95	3.0	4.2	Versicolor
96	2.9	4.2	Versicolor
97	2.9	4.3	Versicolor
98	2.5	3.0	Versicolor
99	2.8	4.1	Versicolor

[100 rows x 3 columns]

## In [55]:

```
var=[]
for i in range(len(iris_data_12)):
    if iris_data_12['variety'].iloc[i]=='Setosa':
        var.append(1)
    else:
        var.append(-1)
iris_data_12['var'] = var
iris_data_12 = iris_data_12.drop(['variety'],axis=1)
print(iris_data_12)
```

```
sepal.width petal.length var
0
           3.5
                       1.4
                              1
           3.0
1
                       1.4
                              1
2
           3.2
                       1.3
                              1
3
           3.1
                       1.5
                              1
4
          3.6
                       1.4
                              1
           . . .
                       95
                       4.2
          3.0
                           -1
96
          2.9
                       4.2
                             -1
97
          2.9
                       4.3
                             -1
98
          2.5
                       3.0
                             -1
99
          2.8
                       4.1
                             -1
```

```
[100 rows x 3 columns]
```

/home/shreja/.local/lib/python3.6/site-packages/ipykernel\_launcher.py:7: S
ettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

```
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy import sys
```

#### In [56]:

```
iris_data_13 = iris_data[iris_data['variety']!='Versicolor']
print(iris_data_13)
```

```
sepal.width petal.length
                                   variety
0
             3.5
                            1.4
                                    Setosa
1
             3.0
                            1.4
                                    Setosa
2
             3.2
                            1.3
                                    Setosa
3
                            1.5
                                    Setosa
             3.1
4
             3.6
                            1.4
                                    Setosa
             . . .
                            . . .
                                        . . .
                            5.2 Virginica
145
             3.0
146
             2.5
                            5.0 Virginica
                            5.2 Virginica
147
             3.0
148
             3.4
                           5.4 Virginica
                            5.1 Virginica
149
             3.0
```

[100 rows x 3 columns]

#### In [57]:

```
var=[]
for i in range(len(iris_data_13)):
    if iris_data_13['variety'].iloc[i]=='Setosa':
        var.append(1)
    else:
        var.append(-1)
iris_data_13['var'] = var
iris_data_13 = iris_data_13.drop(['variety'],axis=1)
print(iris_data_13)
```

	sepal.width	petal.length	var
0	3.5	1.4	1
1	3.0	1.4	1
2	3.2	1.3	1
3	3.1	1.5	1
4	3.6	1.4	1
	• • •		
145	3.0	5.2	-1
146	2.5	5.0	-1
147	3.0	5.2	-1
148	3.4	5.4	-1
149	3.0	5.1	-1

[100 rows x 3 columns]

/home/shreja/.local/lib/python3.6/site-packages/ipykernel\_launcher.py:7: S
ettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html#returning-a-view-versus-a-copy import sys

```
In [58]:
```

```
iris_data_23 = iris_data[iris_data['variety']!='Setosa']
print(iris_data_23)
     sepal.width petal.length
                                   variety
50
             3.2
                           4.7 Versicolor
51
             3.2
                           4.5 Versicolor
52
             3.1
                           4.9 Versicolor
53
             2.3
                           4.0 Versicolor
54
             2.8
                           4.6 Versicolor
             . . .
                           . . .
. .
145
             3.0
                           5.2 Virginica
                          5.0 Virginica
146
             2.5
147
             3.0
                          5.2 Virginica
                           5.4 Virginica
148
             3.4
                           5.1 Virginica
149
             3.0
[100 rows x 3 columns]
In [59]:
var=[]
for i in range(len(iris_data_23)):
    if iris_data_23['variety'].iloc[i]=='Versicolor':
        var.append(1)
    else:
        var.append(-1)
iris_data_23['var'] = var
iris_data_23 = iris_data_23.drop(['variety'],axis=1)
print(iris_data_23)
     sepal.width petal.length var
50
             3.2
                           4.7
                           4.5
51
             3.2
                                  1
52
                           4.9
                                  1
             3.1
                           4.0
53
             2.3
                                  1
54
             2.8
                           4.6
                                  1
             . . .
                           . . .
                           5.2
145
             3.0
                                -1
146
             2.5
                           5.0
                                 -1
                           5.2
                                 -1
147
             3.0
148
             3.4
                           5.4
                                 -1
                           5.1
149
             3.0
                                 -1
[100 rows x 3 columns]
/home/shreja/.local/lib/python3.6/site-packages/ipykernel_launcher.py:7: S
ettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-doc
s/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
```

# **Training For Perceptron**

import sys

## In [60]:

```
data_aug_12 = pd.DataFrame(list(zip(iris_data_12['petal.length'],iris_data_12['sepal.wi
dth'], [1 for i in range(len(iris_data_12))],iris_data_12['var'])), columns=['petal.len
gth','sepal.width','augment','var'])
print(data_aug_12)
```

	petal.length	sepal.width	augment	var
0	1.4	3.5	1	1
1	1.4	3.0	1	1
2	1.3	3.2	1	1
3	1.5	3.1	1	1
4	1.4	3.6	1	1
• •		• • •		
95	4.2	3.0	1	-1
96	4.2	2.9	1	-1
97	4.3	2.9	1	-1
98	3.0	2.5	1	-1
99	4.1	2.8	1	-1

[100 rows x 4 columns]

## In [61]:

```
data_aug_13 = pd.DataFrame(list(zip(iris_data_13['petal.length'],iris_data_13['sepal.wi
dth'], [1 for i in range(len(iris_data_13))],iris_data_13['var'])), columns=['petal.len
gth','sepal.width','augment','var'])
print(data_aug_13)
```

	petal.length	sepal.width	augment	var
0	1.4	3.5	1	1
1	1.4	3.0	1	1
2	1.3	3.2	1	1
3	1.5	3.1	1	1
4	1.4	3.6	1	1
• •	• • •	• • •		• • •
95	5.2	3.0	1	-1
96	5.0	2.5	1	-1
97	5.2	3.0	1	-1
98	5.4	3.4	1	-1
99	5.1	3.0	1	-1

[100 rows x 4 columns]

#### In [62]:

```
data_aug_23 = pd.DataFrame(list(zip(iris_data_23['petal.length'],iris_data_23['sepal.wi
dth'], [1 for i in range(len(iris_data_23))],iris_data_23['var'])), columns=['petal.len
gth','sepal.width','augment','var'])
print(data_aug_23)
```

```
petal.length sepal.width augment var
0
             4.7
                          3.2
                                    1
                                          1
1
             4.5
                          3.2
                                    1
                                          1
2
             4.9
                          3.1
                                    1
                                          1
3
             4.0
                          2.3
                                    1
                                          1
4
            4.6
                         2.8
                                    1
                                          1
             . . .
                          . . .
                                   . . .
95
            5.2
                         3.0
                                    1
                                        -1
                                        -1
96
            5.0
                         2.5
                                    1
97
                         3.0
                                    1 -1
            5.2
98
             5.4
                         3.4
                                   1 -1
                         3.0
                                    1
                                        -1
99
            5.1
```

[100 rows x 4 columns]

Splitting Features and Labels

# In [63]:

```
def split(data_aug):
    data_up = data_aug.copy()
    for i in range(len(data_up)):
        if(data_up['var'][i] == 1):
            data_up['petal.length'][i] = - data_up['petal.length'][i]
            data_up['sepal.width'][i] = -data_up['sepal.width'][i]
            data_up['augment'][i] = -data_up['augment'][i]

# print(data_up)

data_up = data_up.drop(['var'], axis=1)
data_up = np.array(data_up)

# print(data_up)

labels = data_aug['var']
labels = np.array(labels)

# print(Labels)

return (data_up, labels)
```

#### In [64]:

```
data up 12, labels 12 = split(data aug 12)
data_up_13, labels_13 = split(data_aug_13)
data_up_23, labels_23 = split(data_aug_23)
/home/shreja/.local/lib/python3.6/site-packages/ipykernel_launcher.py:5: S
ettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-doc
s/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
/home/shreja/.local/lib/python3.6/site-packages/ipykernel launcher.py:6: S
ettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-doc
s/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
/home/shreja/.local/lib/python3.6/site-packages/ipykernel_launcher.py:7: S
ettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-doc
s/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  import sys
```

## In [65]:

```
def Perceptron_Train(data_up, labels, eta):
   threshold = 5000
    A = np.zeros(len(data_up[0]))
    # print(A)
    for epoch in range(threshold):
        ctr = 0
        for inp, label in zip(data_up, labels):
            summation = np.dot(inp, A)
            if summation <= 0:</pre>
                A += eta*inp
                ctr = 0
            else:
                ctr+=1
        if ctr == len(data up): #Until all input is correctly classified
            break
    print(epoch)
    return A
```

```
In [66]:
x1, x2, b = symbols('x1 x2 1')
P = np.array([x1,x2, b])
print(P)
[x1 x2 1]
In [67]:
eta = 0.01
A_12 = Perceptron_Train(data_up_12, labels_12, eta)
print(A_12)
[ 0.052 -0.041 -0.01 ]
In [68]:
Percep_equation_12 = np.dot(A_12.T, P)
print("Decision Boundary between Setosa and Versicolor: ", Percep_equation_12, ' = 0')
Decision Boundary between Setosa and Versicolor: -0.01*1 + 0.052*x1 - 0.0
41*x2 = 0
In [69]:
eta = 0.01
A_13 = Perceptron_Train(data_up_13, labels_13, eta)
print(A_13)
2
[ 0.032 -0.037 -0.01 ]
In [70]:
Percep_equation_13 = np.dot(A_13.T, P)
print("Decision Boundary between Setosa and Virginica: ", Percep_equation_13, ' = 0')
Decision Boundary between Setosa and Virginica: -0.01*1 + 0.032*x1 - 0.03
7*x2 = 0
In [71]:
eta = 0.01
A_23 = Perceptron_Train(data_up_23, labels_23, eta)
print(A_23)
4999
[ 1.296 -0.837 -3.43 ]
In [72]:
Percep equation 23 = np.dot(A 23.T, P)
print("Decision Boundary between Versicolor and Virginica: ", Percep_equation_23, ' =
 0')
Decision Boundary between Versicolor and Virginica: -3.429999999997*1 +
```

1.29600000000016\*x1 - 0.837\*x2 = 0

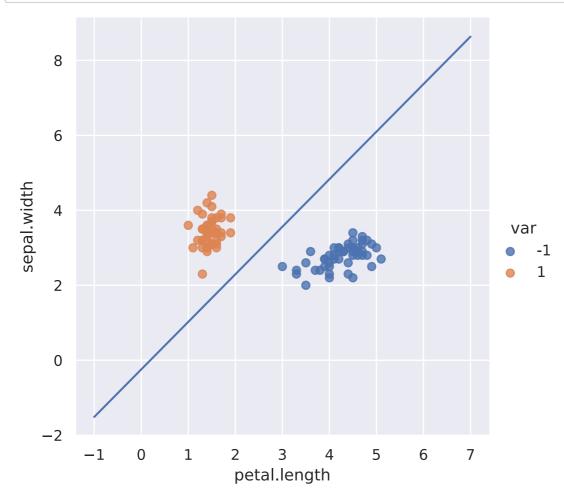
# In [73]:

```
def plot_percep(A, data_aug):
    a = np.linspace(-1,7,100)
    b = -a*A[0]/A[1] - A[2]/A[1]

    sns.set(style="darkgrid")
    sns.lmplot(x='petal.length',y='sepal.width', data=data_aug, fit_reg=False, hue='va'
r', legend=True)
    plt.plot(a,b)
    plt.show()
```

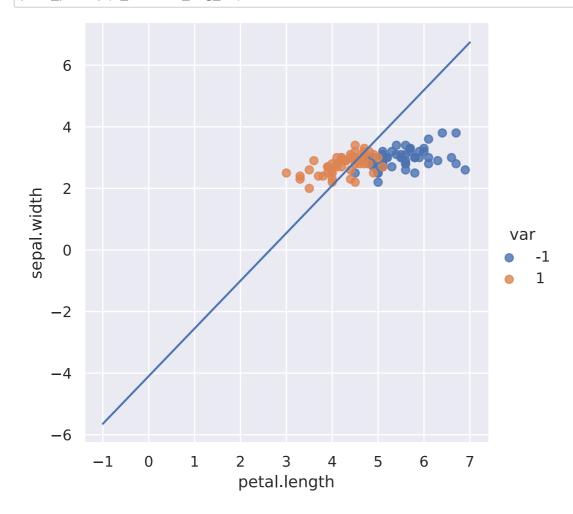
# In [74]:

```
plot_percep(A_12,data_aug_12)
```



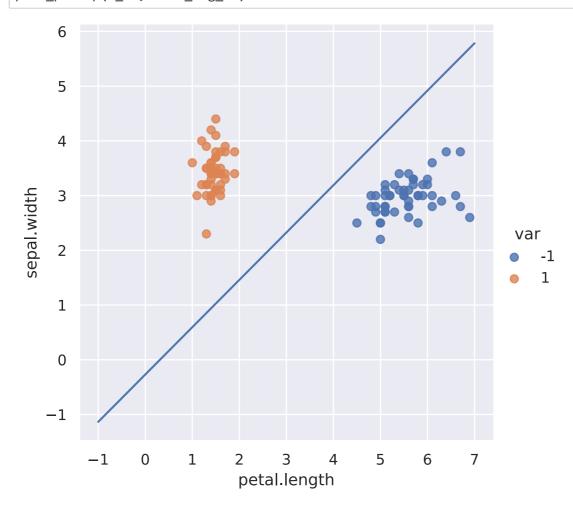
In [75]:

plot\_percep(A\_23,data\_aug\_23)



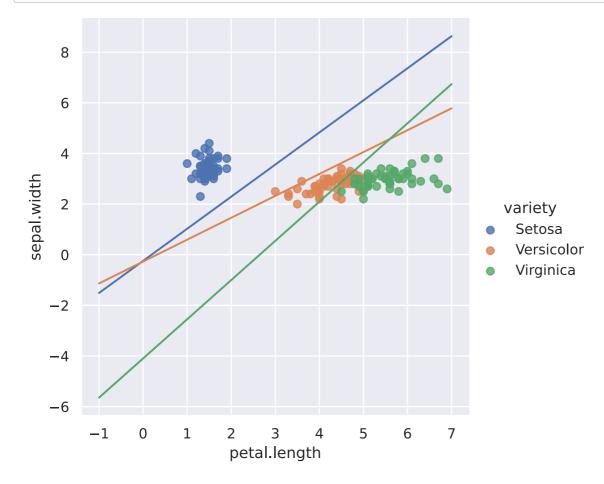
In [76]:

plot\_percep(A\_13, data\_aug\_13)



## In [77]:

```
a = np.linspace(-1,7,100)
b_12 = -a*A_12[0]/A_12[1] - A_12[2]/A_12[1]
b_13 = -a*A_13[0]/A_13[1] - A_13[2]/A_13[1]
b_23 = -a*A_23[0]/A_23[1] - A_23[2]/A_23[1]
sns.set(style="darkgrid")
sns.lmplot(x='petal.length',y='sepal.width', data=iris_data, fit_reg=False, hue='variet
y', legend=True)
# plt.scatter(data_up[:50][0], data_up[:50][1])
# plt.scatter(data_up[50:][0],data_up[50:][1])
plt.plot(a,b_12)
plt.plot(a,b_13)
plt.plot(a,b_23)
plt.show()
```



## **Training with SVM**

#### In [78]:

```
def data_svm_split(iris):
    X = iris[['petal.length','sepal.width']]
    X = np.array(X)
    # print(X)

Y = iris['var']
    Y = np.array(Y).reshape(len(iris),1)
    # print(Y)

return (X,Y)
```

#### In [79]:

```
X_12,Y_12 = data_svm_split(iris_data_12)
X_13,Y_13 = data_svm_split(iris_data_13)
X_23,Y_23 = data_svm_split(iris_data_23)
```

#### In [80]:

```
def SVM_Train(X,Y):
   m = len(X)
    # print(np.dot(Y,Y.T))
    # print(np.dot(X,X.T))
    P = matrix(np.multiply(np.dot(Y, Y.T), np.dot(X, X.T)))
    # print(P)
    q = matrix(np.ones(m) * -1)
    g1 = np.asarray(np.diag(np.ones(m) * -1))
    # g2 = np.asarray(np.diag(np.ones(m)))
    \# G = matrix(np.append(g1, g2, axis=0))
    print(np.array(g1).shape)
    h = matrix(np.zeros(m))
    A = np.reshape((Y.T), (1,m))
    b=[[0]]
    \# b = np.array(b).reshape(m,1)
    P = matrix(P,(m,m),'d') #dense
    A = matrix(A,(1,m),'d')
    g1 = matrix(g1,(m,m),'d')
    b = matrix(b, (1,1), 'd')
    sol = solvers.qp(P, q, g1, h, A, b)
    alpha = np.array(sol['x'])
    ind = (alpha > 1e-4).flatten()
    print(ind)
    W = np.dot(np.transpose(alpha*Y),X)
    print(W)
    for i in range(m):
        if ind[i] == True:
            W0 = 1 - np.dot(X[i], W.T)
            print
            break
    print(W0)
    return (W, W0)
```

#### In [81]:

```
x1, x2 = symbols('x1 x2')
P_SVM = np.array([x1,x2])
print(P_SVM)
```

[x1 x2]

```
In [82]:
```

```
W_{12}, W_{012} = SVM_{Train}(X_{12}, Y_{12})
(100, 100)
                  pcost
                                                                    dcost
                                                                                                                                               pres
                                                                                                                                                                            dres
                                                                                                                    gap
   0: -4.3867e+00 -8.1716e+00
                                                                                                                   3e+02 1e+01
                                                                                                                                                                        2e+00
   1: -2.3798e+00 -2.5258e+00
                                                                                                                  2e+01
                                                                                                                                               1e+00
                                                                                                                                                                          2e-01
   2: -4.3479e-01 -1.7538e+00 2e+00 5e-02
                                                                                                                                                                            6e-03
   3: -6.7595e-01 -1.0544e+00 5e-01 1e-02 1e-03
   4: -8.1435e-01 -1.1260e+00 4e-01 4e-03 5e-04
   5: -1.0085e+00 -1.0271e+00 2e-02
                                                                                                                                               8e-05 1e-05
   6: -1.0251e+00 -1.0253e+00 2e-04 9e-07
                                                                                                                                                                         1e-07
   7: -1.0253e+00 -1.0253e+00 2e-06 9e-09 1e-09
   8: -1.0253e+00 -1.0253e+00 2e-08 9e-11 1e-11
Optimal solution found.
[False False False
   False False False False False False False False False False False
      True False False False False False False False False False False
   False False False False True False False False False False False
   False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False False 
    False False False False False False False False False False False
   False False False False False False False False False False False
   False False False False False False False False False False False
   False False True False]
[[-1.25714286 0.68571428]]
[1.05714287]
In [83]:
SVM_equation_12 = np.dot(W_12[0].T, P_SVM) + W0_12[0]
print("Decision Boundary between Setosa and Versicolor: ", SVM_equation_12, ' = 0')
```

Decision Boundary between Setosa and Versicolor: -1.25714285940527\*x1 +

0.685714284290053\*x2 + 1.05714286628384 = 0

```
In [84]:
```

```
W_{13}, W_{013} = SVM_{Train}(X_{13}, Y_{13})
(100, 100)
                    pcost
                                                                                                                                                             pres
                                                                                                                                                                                            dres
                                                                          dcost
                                                                                                                              gap
   0: -3.3137e+00 -4.9765e+00
                                                                                                                            2e+02 2e+01 1e+00
   1: -1.6126e-01 -8.1531e-01 1e+01 9e-01 8e-02
   2: 1.9106e-02 -6.5205e-01 9e-01 1e-02 1e-03
   3: -1.5487e-01 -2.7974e-01 1e-01 7e-04 6e-05
   4: -2.2572e-01 -2.8972e-01 6e-02 1e-04 1e-05
   5: -2.5983e-01 -2.6459e-01 5e-03 1e-05 8e-07
   6: -2.6415e-01 -2.6421e-01 5e-05 1e-07
                                                                                                                                                                                       9e-09
   7: -2.6420e-01 -2.6420e-01 5e-07 1e-09 9e-11
   8: -2.6420e-01 -2.6420e-01 5e-09 1e-11 9e-13
Optimal solution found.
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   False False False False False False False False False False False
   False False False]
[[-0.68692206 0.23778071]]
[1.49669752]
```

#### In [85]:

```
SVM_equation_13 = np.dot(W_13[0].T, P_SVM) + W0_13[0]
print("Decision Boundary between Setosa and Virginica: ", SVM_equation_13, ' = 0')
```

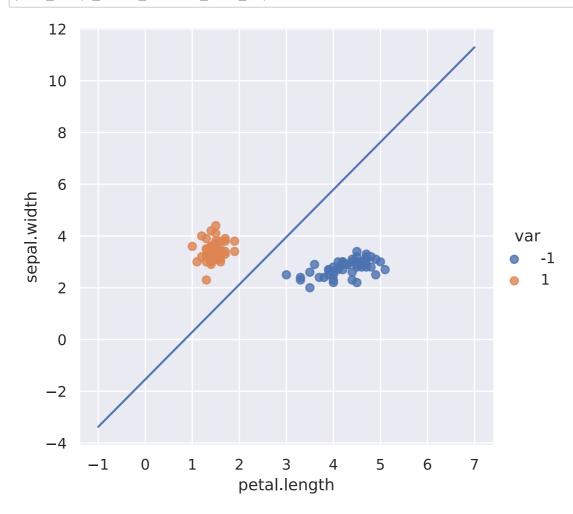
Decision Boundary between Setosa and Virginica: -0.686922063593103\*x1 + 0.237780707125038\*x2 + 1.49669751660177 = 0

```
In [86]:
```

```
W 23, W0 23 = SVM Train(X 23, Y 23)
(100, 100)
                                           dres
    pcost
                                    pres
                 dcost
                            gap
 0: -3.7394e+01 -8.7261e+01
                            5e+02
                                   2e+01
                                           2e+00
 1: -1.0739e+02 -1.7754e+02
                            3e+02
                                   1e+01
                                           2e+00
 2: -5.7593e+02 -8.7585e+02
                            4e+02
                                   1e+01
                                           2e+00
 3: -1.9989e+03 -2.2838e+03
                            3e+02
                                   1e+01
                                           1e+00
 4: -6.2946e+03 -6.9640e+03
                            7e+02
                                   1e+01
                                          1e+00
 5: -3.5722e+04 -3.7828e+04
                            2e+03
                                   1e+01
                                           1e+00
 6: -1.6095e+05 -1.6859e+05
                            8e+03
                                   1e+01
                                           1e+00
 7: -2.7867e+06 -2.8211e+06
                            3e+04
                                   9e+00
                                           1e+00
 8: -1.2047e+08 -1.2120e+08
                            7e+05
                                   9e+00
                                           1e+00
 9: -1.9026e+10 -1.9034e+10
                            8e+06
                                    9e+00
                                           1e+00
                                   9e+00
10: -1.9238e+10 -1.9247e+10 9e+06
                                           1e+00
11: -1.9261e+10 -1.9270e+10
                            9e+06
                                   9e+00
                                           1e+00
12: -3.0216e+10 -3.0229e+10
                            1e+07
                                    9e+00
                                           1e+00
13: -5.7431e+10 -5.7455e+10
                            2e+07
                                   9e+00
                                          1e+00
14: -7.1454e+10 -7.1481e+10 3e+07
                                    9e+00
                                          1e+00
15: -7.9138e+10 -7.9165e+10
                            3e+07
                                    9e+00
                                          1e+00
16: -9.4603e+10 -9.4630e+10 3e+07
                                   9e+00
                                           1e+00
Terminated (singular KKT matrix).
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[[-2.44140625e-04 -7.62939453e-06]]
[1.00117188]
In [87]:
SVM_equation_23 = np.dot(W_23[0].T, P_SVM) + W0_23[0]
print("Decision Boundary between Versicolor and Virginica: ", SVM_equation_23, ' = 0')
Decision Boundary between Versicolor and Virginica: -0.000244140625*x1 -
7.62939453125e-6*x2 + 1.001171875 = 0
In [88]:
def plot_SVM(W,W0, data_aug):
    a = np.linspace(-1,7,100)
    b = -a*W[0][0]/W[0][1] - W0[0]/W[0][1]
    sns.set(style="darkgrid")
    sns.lmplot(x='petal.length',y='sepal.width', data=data_aug, fit_reg=False, hue='va
r', legend=True)
    plt.plot(a,b)
    plt.show()
```

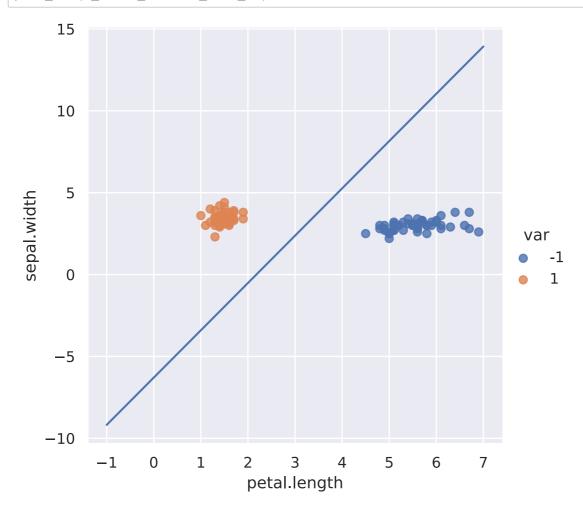
In [89]:

plot\_SVM(W\_12,W0\_12,iris\_data\_12)



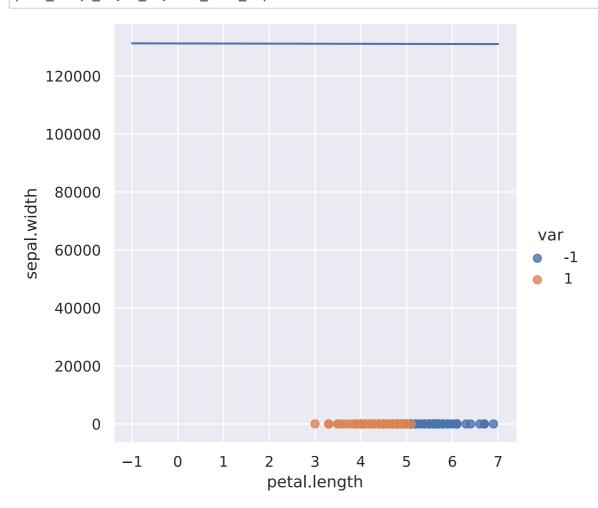
In [90]:

plot\_SVM(W\_13,W0\_13,iris\_data\_13)



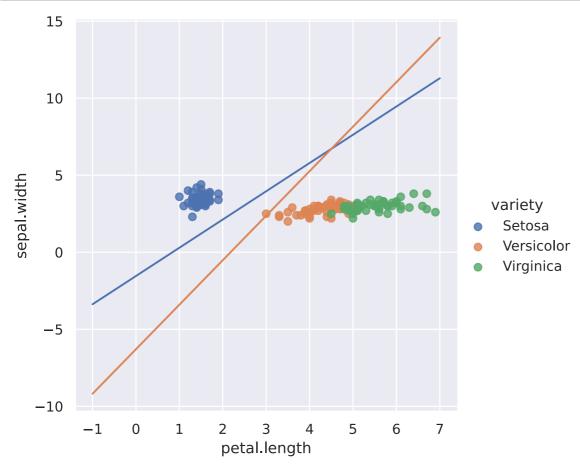
In [91]:

 $\verb"plot_SVM(W_23, W0_23, \verb"iris_data_23")"$ 



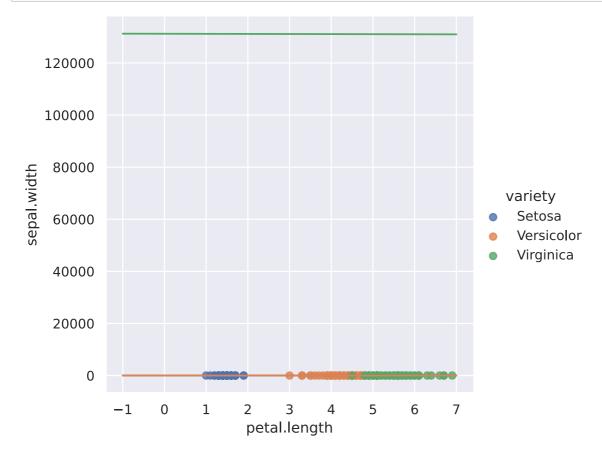
## In [92]:

```
a = np.linspace(-1,7,100)
b_12 = -a*W_12[0][0]/W_12[0][1] - W0_12[0]/W_12[0][1]
b_13 = -a*W_13[0][0]/W_13[0][1] - W0_13[0]/W_13[0][1]
# b_23 = -a*W_23[0][0]/W_23[0][1] - W0_23[0]/W_23[0][1]
sns.set(style="darkgrid")
sns.lmplot(x='petal.length',y='sepal.width', data=iris_data, fit_reg=False, hue='variet y', legend=True)
# plt.scatter(data_up[:50][0], data_up[:50][1])
# plt.scatter(data_up[50:][0],data_up[50:][1])
plt.plot(a,b_12)
plt.plot(a,b_13)
# plt.plot(a,b_23)
plt.show()
```



## In [93]:

```
a = np.linspace(-1,7,100)
b_12 = -a*W_12[0][0]/W_12[0][1] - W0_12[0]/W_12[0][1]
b_13 = -a*W_13[0][0]/W_13[0][1] - W0_13[0]/W_13[0][1]
b_23 = -a*W_23[0][0]/W_23[0][1] - W0_23[0]/W_23[0][1]
sns.set(style="darkgrid")
sns.lmplot(x='petal.length',y='sepal.width', data=iris_data, fit_reg=False, hue='variet
y', legend=True)
# plt.scatter(data_up[:50][0], data_up[:50][1])
# plt.scatter(data_up[50:][0],data_up[50:][1])
plt.plot(a,b 12)
plt.plot(a,b_13)
plt.plot(a,b_23)
plt.show()
```



# In [ ]:

In [ ]: