

$$F(A, B) = \frac{1}{N} \sum \ln (1 + \exp (-y_n (A \cdot Z_n + B)))$$

$$1. \quad \frac{\partial F(A, B)}{\partial A} = \frac{1}{N} \sum \frac{\exp (-y_n (A Z_n + B))}{1 + \exp (-y_n (A Z_n + B))} \cdot (-y_n z_n) = -\frac{1}{N} \sum_{n=1}^N p_n y_n z_n$$

$$\frac{\partial F(A, B)}{\partial B} = -\frac{1}{N} \sum_{n=1}^N p_n y_n$$

$$\nabla F(A, B) = -\frac{1}{N} \sum_{n=1}^N y_n p_n (z_n, 1)$$

$$2. \quad H(F) = \begin{pmatrix} \frac{\partial^2 F}{\partial A^2} & \frac{\partial^2 F}{\partial A \partial B} \\ \frac{\partial^2 F}{\partial B \partial A} & \frac{\partial^2 F}{\partial B^2} \end{pmatrix}$$

$$\frac{\partial^2 F(A, B)}{\partial A^2} = \frac{\partial}{\partial A} \left( -\frac{1}{N} \sum p_n y_n z_n \right) = \frac{1}{N} \sum_{n=1}^N z_n^2 p_n (1 - p_n)$$

$$\frac{\partial^2 F(A, B)}{\partial A \partial B} = \frac{1}{N} \sum_{n=1}^N z_n p_n (1 - p_n)$$

$$\frac{\partial^2 F(A, B)}{\partial B^2} = \frac{1}{N} \sum_{n=1}^N p_n (1 - p_n)$$

$$H(F) = \frac{1}{N} \sum_{n=1}^N p_n (1 - p_n) \begin{bmatrix} z_n^2 & z_n \\ z_n & 1 \end{bmatrix}$$

$$3. \quad \text{when } \gamma \rightarrow \infty \quad K(x, x') = \exp(-\gamma \|x - x'\|^2) \rightarrow 0 \quad (x \neq x')$$

$$\Rightarrow b = 1 \text{ or } -1. \quad \alpha = 0.$$

4. two samples  $\left\{ \begin{array}{l} (X_1, \quad X_1 - X_1^2) \\ (X_2, \quad \underline{X_2 - X_2^2}) \end{array} \right.$

$$E_{OLS} = (W_1 X_1 + W_0 - X_1 + X_1^2)^2 + (W_1 X_2 + W_0 - X_2 + X_2^2)^2$$

$$\frac{\partial E_{OLS}}{\partial W_1} = 2(W_1 X_1 + W_0 - X_1 + X_1^2) X_1 + 2(W_1 X_2 + W_0 - X_2 + X_2^2) X_2 = 0$$

$$\Rightarrow W_1 = (1 - X_1 - X_2)$$

$$\frac{\partial E_{OLS}}{\partial W_0} = 2(W_1 X_1 + W_0 - X_1 + X_1^2) + 2(W_1 X_2 + W_0 - X_2 + X_2^2) = 0$$

$$\Rightarrow W_0 = +X_1 X_2$$

$$\Rightarrow g(x) = (1 - X_1 - X_2) X + X_1 \cdot X_2$$

$$\bar{g}(x) = E[g(x)] = \frac{1}{4}$$

5.  $\sum_n (y_n - W^T X_n)^2 = \left( \sqrt{\sum_n} y_n - W^T \sqrt{\sum_n} X_n \right)^2$

$$\{(\tilde{X}_n, \tilde{y}_n)\}_{n=1}^N = \{(\sqrt{\sum_n} X_n, \sqrt{\sum_n} y_n)\}_{n=1}^N$$

6. error rate = 0.22 =  $\epsilon t$ .

$$\frac{\mu_+^{(2)}}{\mu_-^{(2)}} = \frac{\epsilon t}{1 - \epsilon t} = \frac{22}{78} = \frac{11}{39}$$



7. First, all positive and negative are two decision steps (D-S)

For each dimension, there are  $2 \times M$  interval between  $-M$  and  $M$ , each could be  $+1$  or  $-1$ .  
total  $d$  dimensions,  $\Rightarrow$  total  $4dM + 2$  DSs.

8.  $K_{ds}(x, x')$  is inner product of  $\phi_{ds}(x)$

$$= \frac{4dM + 2}{2} - \sum_{n=1}^{|B|} |x_n - x'_n|$$

$$\text{if } x = x' \Rightarrow K(x, x') = 4dM + 2.$$

```
In [164]: import pandas as pd
import numpy as np
from sklearn.linear_model import Ridge
from sklearn.utils import resample
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```
In [165]: df = np.loadtxt('hw2_1ssvm_all.dat')
```

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In [166]: dftrain = df[0:400]
dftest = df[400:]
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```
In [167]: Lambda = [0.05, 0.5, 5, 50, 500]
```

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In [ ]:
```

```
In [168]: Ein = []
Eout = []
for l in Lambda:
    clf = Ridge(alpha=l)
    clf.fit(dftrain[:,0:-1], dftrain[:,-1])
    y_pred_in = (clf.predict(dftrain[:,0:-1])>0)*2-1
    y_pred_out = (clf.predict(dftest[:,0:-1])>0)*2-1
    Ein = Ein + [sum(y_pred_in != dftrain[:,-1])/len(y_pred_in)]
    Eout = Eout + [sum(y_pred_out != dftest[:,-1])/len(y_pred_out)]
```

```
In [169]: Ein
```

```
Out[169]: [0.3175, 0.3175, 0.3175, 0.32, 0.3225]
```

```
In [170]: Eout
```

```
Out[170]: [0.36, 0.36, 0.36, 0.37, 0.37]
```

With  $\lambda$  increases, Ein and Eout increase.

```
In [171]: Ein = []
Eout = []
for l in Lambda:
    y_bag_out = np.zeros(100)
    y_bag_in = np.zeros(400)
    for t in range(250):
        df_bs = resample(dftrain,n_samples=400,random_state=t)
        clf = Ridge(alpha=l)
        clf.fit(df_bs[:,0:-1], df_bs[:,-1])
        y_bag_out = y_bag_out + (( clf.predict(dftest[:,0:-1]) > 0 ) * 2 - 1)
        y_bag_in = y_bag_in + (( clf.predict(dftrain[:,0:-1]) > 0 ) * 2 - 1)

    Ein = Ein + [sum( ((y_bag_in>0)*2-1) != dftrain[:,-1] ) / len(y_bag_in)]
    Eout = Eout + [sum( ((y_bag_out>0)*2-1) != dftest[:,-1] ) / len(y_bag_out)]
```

```
In [172]: Ein
```

```
Out[172]: [0.3175, 0.3175, 0.3175, 0.3175, 0.3225]
```

In [173]: Eout

Out[173]: [0.36, 0.36, 0.36, 0.36, 0.37]

With  $\lambda$  increases, Ein and Eout increase. And there is no improvement after using bootstrapping.

In [ ]:

In [ ]:

In [ ]: