

CSIE 5432/5433 HW5 PDF

YiWenLai

TOTAL POINTS

30 / 0

QUESTION 1

1 Problem 6 20 / 0

✓ + 20 pts Correct

+ 10 pts (1) Correct answer with one mistake

(2) Correct answer but some descriptions not clear enough.

+ 0 pts (1) Choose "I have no idea" option.

(2) Correct answer but missing some steps.

- 10 pts Wrong answer

QUESTION 2

2 Problem 12 10 / 0

+ 20 pts Correct answer.

✓ + 10 pts 1 mistake.

+ 0 pts 2 mistakes or 2 unclear places.

+ 0 pts Choose "I have no idea" option.

- 10 pts Wrong answer.

① without explaining why is "min"

QUESTION 3

3 Problem 15 0 / 0

✓ + 0 pts Good submission or choose "I have no idea" option

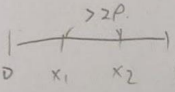
- 10 pts Do not select page or select wrong page

- 20 pts No submission

- 20 pts Only program output or terminal commands

- 20 pts submit a url of code

4. if x_1, x_2 are both ^{with} same sign, the expected dictomomies are 2.
 So we just consider the other two cases with different signs.

 , $p \in [0, 0.5]$, $|x_2 - x_1| > 2p$.

$$\int_0^{1-2p} \int_{x_1+2p}^1 dx_2 \cdot dx_1 = \int_0^{1-2p} (1 - (2p + x_1)) dx_1, \text{ if } x_1 + 2p < 1, \text{ then } |x_2 - x_1| > 0.$$

$$= \int_0^{1-2p} (1-2p) - x_1 dx_1 = (1-2p) \Big|_0^{1-2p} - \frac{1}{2} (1-2p)^2 = \frac{1}{2}$$

$$> \cdot \left(\frac{1}{2} (1-2p)^2 \right) = (1-2p)^2, \text{ for } x_2 > x_1, \text{ same as } x_1 > x_2,$$

$$\therefore \text{ overall, } 2 + 2 \cdot (1-2p)^2$$

5. c

6. e

$$3. \mathcal{L}(b, w, \alpha) = \frac{1}{2} w^T w + \sum [\gamma_n = 1] \alpha_n (P^+ - \gamma_n w^T x_n + b) + \sum [\gamma_n = -1] \alpha_n (P^- - \gamma_n w^T x_n + b)$$

$$\frac{\partial}{\partial b} = 0, \text{ problem} = \max \left(\min \frac{1}{2} w^T w + \sum [\gamma_n = 1] \alpha_n (P^+ - \gamma_n w^T x_n) + \sum [\gamma_n = -1] \alpha_n (P^- - \gamma_n w^T x_n) \right)$$

$$\frac{\partial}{\partial w_i} = 0, \quad w = \sum \alpha_n [\gamma_n = 1] x_n + \sum \alpha_n [\gamma_n = -1] x_n = \sum \alpha_n \gamma_n x_n.$$

$$\text{problem} = \max \sum P^+ [\gamma_n = 1] \alpha_n + \sum P^- [\gamma_n = -1] \alpha_n - \frac{1}{2} w^T w$$

$\downarrow \times (-1)$

$$= \min \frac{1}{2} w^T w - \left(\sum_{n=1}^N P_+ [\gamma_n = 1] \alpha_n + \sum_{n=1}^N P_- [\gamma_n = -1] \alpha_n \right) \quad \times$$

6. for original hard margin, $\sum \alpha_n^* \gamma_n = 0$

uneven-margin, $\sum \alpha_n \gamma_n = 0$

$$\sum \alpha_n^* \gamma_n = \sum_{\gamma_n=+1} \alpha_n^* - \sum_{\gamma_n=-1} \alpha_n^* = 0; \quad \sum_{\gamma_n=+1} \alpha_n - \sum_{\gamma_n=-1} \alpha_n = p, \quad \sum_{\gamma_n=+1} \alpha_n = \sum_{\gamma_n=-1} \alpha_n$$

$$P_+ \sum_{\gamma_n=+1} \alpha_n + P_- \sum_{\gamma_n=-1} \alpha_n = (P_+ + P_-) \sum_{\gamma_n=+1} \alpha_n = (P_+ + P_-) \sum_{\gamma_n=-1} \alpha_n = \left(\frac{P_+ + P_-}{2} \right) \sum_{n=1}^N \alpha_n$$

Under gradient constraint of optimal solution

$Q\alpha + P = 0$, for un-even case

$Q\alpha^* + i = 0$, for original case.

\downarrow

$Q \left(\frac{P_+ + P_-}{2} \right) \alpha^* + \frac{(P_+ + P_-)}{2} = 0$ is also a solution!

$$\frac{P_+ + P_-}{2} \alpha^*$$

7. d
8. c
9. d
10. c

1 Problem 6 20 / 0

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- 10 pts Wrong answer

7.

- d. $\log_2 K(x, x')$ may be less than 0 when $0 < K(x, x') < 1$, which may cause the element of matrix non-positive. not positive definite matrix $ZZ^T \rightarrow$ not valid kernel ✖

8.

$$\begin{aligned} \|\phi(x) - \phi(x')\|^2 &= \phi(x)^T \phi(x) - 2\phi(x)^T \phi(x') + \phi(x')^T \phi(x') \\ &= K(x, x) + K(x', x') - 2K(x, x') \\ &= 1 + 1 - 2K(x, x') \leq 2 \quad \text{✖} \end{aligned}$$

10.

$$w_{t+1} = \sum_{n=1}^N \alpha_{t+1} \phi(x_n) = w_t + y_n(t) \phi(x_n(t)) = \left(\sum_{n=1}^N \alpha_t \phi(x_n) \right) + y_n(t) \phi(x_n)$$

(compare coefficient of each $\phi(x)$, $\alpha_{t+1} = \alpha_t$ except $\phi(x_n)$, whose coefficient is $\alpha_{t,n(t)} + y_n(t)$) ✖

9.

problem can be viewed as:

$$h(x_m) = \text{sign} \left(\sum_{n=1}^N y_n K(x_m, x_n) \right) = y_m, \quad \forall m = 1, 2, \dots, N, \quad \text{E in } 0$$

$$\sum_{n=1}^N y_n K(x_m, x_n) = \sum_{\substack{n=1 \\ n \neq m}}^N y_n \exp(-r(x_m - x_n)^2) + y_m$$

$$\left| \sum_{\substack{n=1 \\ n \neq m}}^N y_n \exp(-r(x_m - x_n)^2) \right| < 1, \text{ will maintain constraint.}$$

$$-1 < \sum y_n \exp(-r(x_m - x_n)^2) \leq \sum y_n \exp(-r\epsilon^2) < 1$$

$$-1 < -(N-1) \exp(-r\epsilon^2) < \sum y_n \exp(-r\epsilon^2) < (N-1) \exp(-r\epsilon^2) < 1, \text{ if } r \text{ is large enough.}$$

$$\therefore \exp(-r\epsilon^2) < \frac{1}{N-1} \text{ for both side } -r\epsilon^2 < \ln\left(\frac{1}{N-1}\right) \quad r > \frac{\ln(N-1)}{\epsilon^2} \quad \text{✖}$$

11. a
12. b
13. e
14. e

$$11. \quad w_t^T \phi(x) = \left(\sum_{n=1}^N \alpha_n \phi(x_n) \right)^T \phi(x) = \sum_{n=1}^N \alpha_n \phi(x_n)^T \phi(x) = \sum_{n=1}^N \alpha_n k(x_n, x) \quad \times$$

12. $\alpha_n = C, \forall n$, means that $1 - \epsilon_n - y_n(w^T z_n + b) = 0$
 $\epsilon_n \geq 0, 1 - y_n(w^T z_n + b) \geq 0, 1 - y_n w^T z_n \geq y_n b$.
 $y_n = 1 \Rightarrow 1 - w^T z_n \geq b$ to find the upper bound $b \leq 1 - \sum_{n=1}^N y_n \alpha_n k(x_n, x_n)$ is upper bound \times
 $y_n = -1 < 0, -1 - w^T z_n \leq b$.

13. $\max \frac{1}{2} w^T w + C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (1 - \epsilon_n - y_n(w^T z_n + b))$

$$\left(\frac{\partial}{\partial b} \right)$$

$$\max \frac{1}{2} w^T w + C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (1 - \epsilon_n - y_n w^T z_n)$$

$$\left(\frac{\partial}{\partial w} \right), w = \sum_{n=1}^N y_n \alpha_n z_n$$

$$\max C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (-\epsilon_n) - \frac{1}{2} \left\| \sum_{n=1}^N \alpha_n y_n z_n \right\|^2 + \sum_{n=1}^N \alpha_n$$

$$\frac{\partial}{\partial \epsilon_n} \downarrow$$

$$2C \sum \epsilon_n - \sum \alpha_n = 0, \alpha_n = 2C \epsilon_n$$

$$\max -\frac{1}{2} \left\| \sum \alpha_n y_n z_n \right\|^2 - C \sum \epsilon_n^2 + \sum_{n=1}^N \alpha_n$$

$$\min \frac{1}{2} \sum_n \sum_m \alpha_n \alpha_m y_n y_m z(n)^T z(m) + \frac{1}{2} \left[\sum_n \frac{(\alpha_n)^2}{2C} \right] + \sum_{n=1}^N \alpha_n$$

$$= \frac{1}{2} \sum_n \sum_m \alpha_n \alpha_m y_n y_m \left(k(x_n, x_m) + \frac{1}{2C} [n=m] \right) (y_n^2 = 1) + \sum_{n=1}^N \alpha_n \quad \times$$

14.

$$\text{as above show, } \frac{\alpha^*}{2C} = \epsilon^* \quad \times$$

15. d

a. result, 8.457084298367683

2 Problem 12 10 / 0

+ **20 pts** Correct answer.

✓ + **10 pts** 1 mistake.

+ **0 pts** 2 mistakes or 2 unclear places.

+ **0 pts** Choose "I have no idea" option.

- **10 pts** Wrong answer.

1 without explaining why is "min"

$$11. \quad w_t^T \phi(x) = \left(\sum_{n=1}^N \alpha_n \phi(x_n) \right)^T \phi(x) = \sum_{n=1}^N \alpha_n \phi(x_n)^T \phi(x) = \sum_{n=1}^N \alpha_n k(x_n, x) \quad \times$$

12. $\alpha_n = C, \forall n$, means that $1 - \epsilon_n - y_n(w^T z_n + b) = 0$
 $\epsilon_n \geq 0, 1 - y_n(w^T z_n + b) \geq 0, 1 - y_n w^T z_n \geq y_n b$.
 $y_n = 1 \Rightarrow 1 - w^T z_n \geq b$ to find the upper bound b is upper bound \times
 $y_n = -1 < 0, -1 - w^T z_n \leq b$.

13. $\max \frac{1}{2} w^T w + C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (1 - \epsilon_n - y_n(w^T z_n + b))$

$$\left(\frac{\partial}{\partial b} \right)$$

$$\max \frac{1}{2} w^T w + C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (1 - \epsilon_n - y_n w^T z_n)$$

$$\left(\frac{\partial}{\partial w} \right), w = \sum_{n=1}^N y_n \alpha_n z_n$$

$$\max C \sum_{n=1}^N \epsilon_n^2 + \sum_{n=1}^N \alpha_n (-\epsilon_n) - \frac{1}{2} \left\| \sum_{n=1}^N \alpha_n y_n z_n \right\|^2 + \sum_{n=1}^N \alpha_n$$

$$\frac{\partial}{\partial \epsilon_n} \downarrow$$

$$2C \sum \epsilon_n - \sum \alpha_n = 0, \alpha_n = 2C \epsilon_n$$

$$\max -\frac{1}{2} \left\| \sum \alpha_n y_n z_n \right\|^2 - C \sum \epsilon_n^2 + \sum_{n=1}^N \alpha_n$$

$$\min \frac{1}{2} \sum_n \sum_m \alpha_n \alpha_m y_n y_m z(n)^T z(m) + \frac{1}{2} \left[\sum_n \frac{(\alpha_n)^2}{2C} \right] + \sum_{n=1}^N \alpha_n$$

$$= \frac{1}{2} \sum_n \sum_m \alpha_n \alpha_m y_n y_m \left(k(x_n, x_m) + \frac{1}{2C} [n=m] \right) (y_n^2 = 1) + \sum_{n=1}^N \alpha_n \quad \times$$

14.

$$\text{as above show, } \frac{\alpha^*}{2C} = \epsilon^* \quad \times$$

15. d

a. result, 8.457084298367683


```

yiwenlai@YiWens-MBP ~/Desktop/html_hw5/libsvm-3.24/python → INSERT python3 train.py
.....*.....*
optimization finished, #iter = 22874
nu = 0.108695
obj = -4784.881503, rho = 3.623003
nSV = 500, nBSV = 466
Total nSV = 500
Accuracy = 95.8061% (4249/4435) (classification)
8.457084298367683

```

b. Code

```

y, x = svm_read_problem('hw5_train')
for i in range(len(y)):
    if y[i] == 3.0:
        y[i] = 1
    else:
        y[i] = -1
m = svm_train(y, x, '-c 10 -s 0 -t 0')
p_label, p_acc, p_val = svm_predict(y, x, m)
support_vectors = m.get_SV()
support_vector_coefficients = m.get_sv_coef()
w = []
for i in range(36):
    wi = 0
    for j in range(len(support_vector_coefficients)):
        if i+1 in support_vectors[j]:
            wi +=
support_vector_coefficients[j][0]*support_vectors[j]
[i+1]
    w.append(wi)
w_2 = [x*x for x in w]
print(math.sqrt(sum(w_2)))

```

16. b

a. result is presented in Q17

17. c

a. result for Q16,17

```

yiwenlai@YiWens-MBP ~/Desktop/html_hw5/libsvm-3.24/python → INSERT python3 train.py
Class: 1 Number of support vectors: 145 Accuracy: 99.93235625704622
Class: 2 Number of support vectors: 87 Accuracy: 100.0
Class: 3 Number of support vectors: 433 Accuracy: 97.76775648252537
Class: 4 Number of support vectors: 712 Accuracy: 95.98647125140924
Class: 5 Number of support vectors: 259 Accuracy: 99.32356257046223

```

b. Code for Q16,17

3 Problem 15 0 / 0

✓ + **0 pts** Good submission or choose "I have no idea" option

- **10 pts** Do not select page or select wrong page
- **20 pts** No submission
- **20 pts** Only program output or terminal commands
- **20 pts** submit a url of code