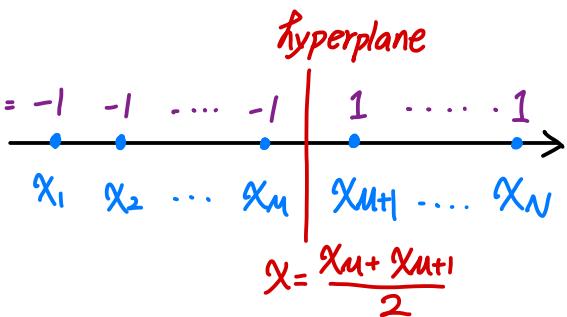


PI.A 考慮 1D



$$g_{\text{SVM}} = \text{sign}(Wx + b)$$

顯然當 $W=1$, $b = -\frac{x_m + x_{m+1}}{2}$. 所有 $x_1 \sim x_n$ 皆能正確分類

$$\underline{g_{\text{SVM}} = \text{sign}(X_i - \frac{x_m + x_{m+1}}{2})}, \text{故选 [a]}$$

P2.D

證明 $\|w\|^{-1}$ 是 margin distance

By L10, P10 的距離公式：distance = $\frac{1}{\|w\|} \cdot y_n (w^T z_n + b)$

By KKT condition: $b = y_n - w^T z_n$ 代入上式得：

$$\Rightarrow \text{distance} = \frac{1}{\|w\|} \cdot (w^T z_n + y_n - w^T z_n) y_n = \frac{y_n^2}{\|w\|} = \frac{1}{\|w\|} \# \text{故选 [3]}$$

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

By L10, P31 的 KKT condition

$$\max_{\text{all } \alpha_n > 0, \sum y_n \alpha_n = 0, W = \sum \alpha_n y_n z_n} -\frac{1}{2} \left\| \sum_{n=1}^N \alpha_n y_n z_n \right\|^2 + \sum_{n=1}^N \alpha_n$$

條件① 條件② 代入條件③ ↑

$$\Rightarrow -\frac{1}{2} \|W\|^2 + \sum_{n=1}^N \alpha_n$$

$$\Rightarrow \frac{1}{2} W^T W + \sum_{n=1}^N \alpha_n - W^T \boxed{W}$$

條件②
代入 $W = \sum \alpha_n y_n z_n$

$$\Rightarrow \frac{1}{2} W^T W + \sum_{n=1}^N \alpha_n - \sum_{n=1}^N \alpha_n y_n W^T z_n$$

$$\Rightarrow \frac{1}{2} W^T W + \sum_{n=1}^N \alpha_n \left(1 - \sum_{n=1}^N y_n W^T z_n \right)$$

代入條件②

$$\Rightarrow \frac{1}{2} W^T W + \sum_{n=1}^N \alpha_n \left(1 - \sum_{n=1}^N y_n (w^T z_n + b) \right)$$

代入 complementary slackness: $\alpha_n (1 - y_n (w^T z_n + b)) = 0$

由於只考慮 SV, 故 $1 - y_n (w^T z_n + b) = 0$

$$\Rightarrow \frac{1}{2} \|W\|^2$$

由前頁推導可知：

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

$$\Rightarrow \underbrace{\left[2 \sum_{n=1}^N \alpha_n - \left\| \sum_{n=1}^N \alpha_n y_n z_n \right\|^2 \right]^{\frac{1}{2}}}_{\#} = \|w\|^{-1} \xrightarrow{\text{式①}} \text{margin distance 故选 [6]}$$

證明 $\left(\sum_{n=1}^N \alpha_n \right)^{\frac{1}{2}}$

根據前面推導的式①

$$\left[2 \sum_{n=1}^N \alpha_n - \left\| \sum_{n=1}^N \alpha_n y_n z_n \right\|^2 \right]^{\frac{1}{2}} = \|w\|$$

$$\Rightarrow 2 \sum_{n=1}^N \alpha_n - \|w\|^2 = \|w\|^2$$

$$\Rightarrow \sum_{n=1}^N \alpha_n = \|w\|^2$$

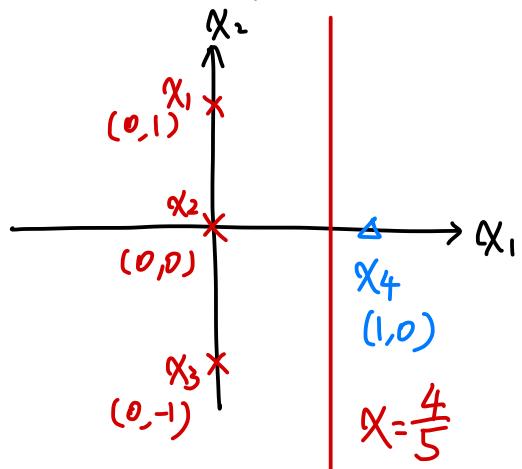
$$\Rightarrow \underbrace{\left(\sum_{n=1}^N \alpha_n \right)^{\frac{1}{2}}}_{\#} = \|w\|^{-1} = \text{margin distance}$$

P3 C 代入選項C，限制式變成 $x_n = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

if $y_n > 0$, $(w^T x_n + b) \geq 1 \Rightarrow 5x_1 - 4 \geq 1 \Rightarrow x_1 \geq 1$

if $y_n < 0$, $-(w^T x_n + b) \geq -1 \Rightarrow -5x_1 + 4 \geq -1 \Rightarrow x_1 \leq 1$.

顯然滿足題目給的 $x_1 \sim x_4$



P4.C By P3 的 primal SVM: $\min_{w,b} \frac{1}{2} w^T w$. s.t. $\begin{cases} (w^T x_n + b) \geq 1, & y_n > 0 \\ -(w^T x_n + b) \geq -\rho, & y_n < 0 \end{cases}$

By Slide 110, P.26

$$\mathcal{L}(b, w, \alpha) = \frac{1}{2} w^T w + \sum_{n=1}^N \alpha_n (1 - w^T x_n - b) \cdot [y_n = 1]$$

$$+ \sum_{n=1}^N \alpha_n (\rho - w^T x_n - b) \cdot [y_n = -1]$$

$$\frac{\partial \mathcal{L}(b, w, \alpha)}{\partial b} = -\sum_{n=1}^N \alpha_n [y_n = 1] + \sum_{n=1}^N \alpha_n [y_n = -1] = 0 \quad \text{...代入上式, } b \text{ 被消掉}$$

condition ①

$$\Rightarrow \min_{b,w} \frac{1}{2} w^T w + \sum_{n=1}^N \alpha_n (1 - w^T x_n) [y_n = 1] + \sum_{n=1}^N \alpha_n (\rho - w^T x_n) [y_n = -1]$$

$$\frac{\partial \mathcal{L}(b, w, \alpha)}{\partial w_i} = w_i - \sum_{n=1}^N \alpha_n x_n [y_n = 1] + \sum_{n=1}^N \alpha_n x_n [y_n = -1] = 0$$

$$\Rightarrow w = \sum_{n=1}^N \alpha_n x_n [y_n = 1] - \sum_{n=1}^N \alpha_n x_n [y_n = -1] = 0$$

condition ②

代入 $\mathcal{L}(b, w, \alpha)$

$\max_{b,w} (\min_{b,w} \mathcal{L}(b, w, \alpha)) \leftarrow$ 代入 condition ①, ②, 為求過程簡化下面不特別寫出限制條件

$$\Rightarrow \max \left[\frac{1}{2} w^T w + \left(\sum_{n=1}^N \alpha_n [y_n = 1] - w^T w \right) + \sum_{n=1}^N \alpha_n \rho [y_n = -1] - w^T w \right]$$

$$\Rightarrow \max \left[-\frac{1}{2} w^T w + \sum_{n=1}^N \alpha_n [y_n = 1] + \sum_{n=1}^N \alpha_n \rho [y_n = -1] \right]$$

$$\Rightarrow \min \left[+\frac{1}{2} w^T w - \sum_{n=1}^N \alpha_n [y_n = 1] - \sum_{n=1}^N \alpha_n \rho [y_n = -1] \right]$$

後續

$$\Rightarrow \min \left[+\frac{1}{2} W^T W - \sum_{n=1}^N \alpha_n \mathbb{I}[y_n = 1] \mathbb{J} - \sum_{n=1}^N \alpha_n \rho_- \mathbb{I}[y_n = -1] \mathbb{J} \right]$$

$$\Rightarrow \min \left[\frac{1}{2} \|\alpha_n y_n x_n\|^2 - \sum_{n=1}^N \alpha_n \mathbb{I}[y_n = 1] \mathbb{J} - \sum_{n=1}^N \alpha_n \rho_- \mathbb{I}[y_n = -1] \mathbb{J} \right]$$

對應題目選項[C]

P5.D 若只考慮 Support Vector, 分二種情況 $y_n = 1$ or $y_m = -1$

$$\text{even SVM} \Rightarrow \begin{cases} w^T x_n + b = 1, \text{ if } y_n = 1 & \dots (1) \\ w^T x_m + b = -1, \text{ if } y_m = -1 & \dots (2) \end{cases} \Rightarrow (1) - (2) \Rightarrow w^T(x_n - x_m) = 2 \dots (5)$$

$$\text{uneven SVM} \Rightarrow \begin{cases} w^T x_n + b' = 1, \text{ if } y_n = 1 & \dots (3) \\ w^T x_m + b' = -\rho_-, \text{ if } y_m = -1 & \dots (4) \end{cases} \Rightarrow (3) - (4) \Rightarrow w^T(x_n - x_m) = 1 + \rho_- \dots (6)$$

$$\frac{(5)}{(6)} \Rightarrow \frac{w^T(x_n - x_m)}{w^T(x_n - x_m)} = \frac{2}{1 + \rho_-} \Rightarrow \underline{w'} = \frac{1 + \rho_-}{2} \underline{w}$$

By KKT condition: $w = \sum_{n=1}^N \alpha_n y_n x_n$ 代入上式

$$\Rightarrow \sum_{n=1}^N \alpha_n y_n \cancel{x_n} = \frac{1 + \rho_-}{2} \left(\sum_{n=1}^N \alpha_n y_n \cancel{x_n} \right)$$

$$\Rightarrow \sum_{n=1}^N \alpha_n = \frac{1 + \rho_-}{2} \sum_{n=1}^N \alpha_n$$

故选 [D]

P6, A $K(x, x') = (x^T x')^Q$, 代表 $\Phi(x)$ 只留最高次項

$$\Phi(x) = (x_1^Q, x_2^Q, \dots, x_d^Q, x_1 x_2^{Q-1}, x_1 x_3^{Q-1}, \dots, x_1 x_d^{Q-1}, \dots)$$

我們可以想像有 Q 個相同的球，要放進 d 個箱子裡，一共有幾種不同的放法？

$$\Rightarrow |0|0||0 \dots 0||$$

- 共有 Q 個球， $d-1$ 個 bar 做排列 $\Rightarrow \frac{(Q+d-1)!}{Q!(d-1)!} \Rightarrow \underline{\binom{Q+d-1}{Q}}$ 故選 [A]
此為不全相異排列

P.D

$$\text{By Slide L11, P7: } K_2(x, x') = 1 + 2x^T x' + (x^T x')^2 \\ = (1 + x^T x')^2$$

where $\Phi_2(x) = (1, \sqrt{2}x_1, \sqrt{2}x_2, \dots, \sqrt{2}x_d, x_1^2, x_2^2, \dots, x_d^2)$

令 $x = (1, 0, 0, \dots, 0)$, $x' = (-1, 0, 0, \dots, 0)$, 此情形會出現 upper bound.

$$\Phi(x) - \Phi(x') = (\cancel{1}, \cancel{\sqrt{2}x_1}, \cancel{\sqrt{2}x_2}, \dots, \cancel{\sqrt{2}x_d}, \cancel{x_1^2}, \cancel{x_2^2}, \dots, \cancel{x_d^2}) \\ - (\cancel{1}, \cancel{\sqrt{2}x'_1}, \cancel{\sqrt{2}x'_2}, \dots, \cancel{\sqrt{2}x'_d}, \cancel{x'_1^2}, \cancel{x'_2^2}, \dots, \cancel{x'_d^2}) \\ \Rightarrow \sqrt{2} - (-\sqrt{2}) = 2\sqrt{2}$$

$$\Rightarrow \|\Phi(x) - \Phi(x')\|^2 = (2\sqrt{2})^2 = \underline{8}_{\#}$$

P8.C

PLA 更新公式： $W_{t+1} \leftarrow W_t + \underline{y}_{n(t)} \phi(X_{n(t)})$

所以 $\alpha_{t+1}[n(t)] = \underline{y}_{n(t)}$, 才能達到更新的效果

P9. A

Primal SVM 根據題目：

$$\min_{w, b, \xi} \frac{1}{2} w^T w + \sum_{n=1}^N \mathcal{U}_n \cdot \xi_n, \text{ s.t. } y_n (w^T \Phi(x_n) + b) \geq 1 - \xi_n$$

$$\xi \geq 0, n = 1, \dots, N$$

By Slide L11. P27 推導成 dual

$$\mathcal{L}(b, w, \xi, \alpha, \beta, \mathcal{U}_n) = \frac{1}{2} w^T w + \sum_{n=1}^N \mathcal{U}_n \cdot \xi_n + \sum_{n=1}^N \alpha_n \cdot (1 - \xi_n - y_n (w^T z_n + b))$$

$$+ \sum_{n=1}^N \beta_n \cdot (-\xi_n)$$

$$\frac{\partial \mathcal{L}}{\partial \xi_n} = 0 \Rightarrow \underline{\mathcal{U}_n - \alpha_n - \beta_n = 0} \quad \text{condition}$$

$$\max_{b, w, \xi} \left(\min_{\mathcal{U}, \alpha, \beta} \mathcal{L}(b, w, \xi, \alpha, \beta, \mathcal{U}_n) \right)$$

$$\Rightarrow \max_{b, w, \xi} \left(\min_{\mathcal{U}, \alpha, \beta} \frac{1}{2} w^T w + \sum_{n=1}^N \alpha_n (1 - y_n (w^T z_n + b)) \right) + \sum_{n=1}^N (\mathcal{U}_n - \alpha_n - \beta_n) \cdot \xi_n.$$

$\beta_n = \mathcal{U}_n - \alpha_n$

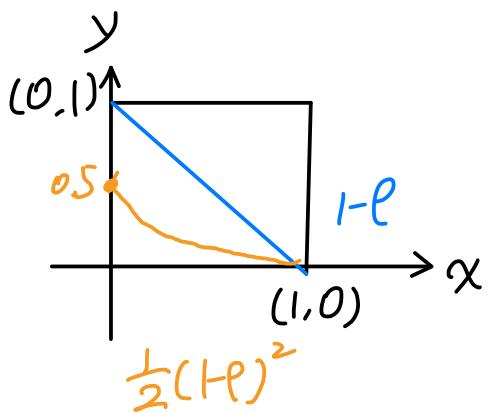
\mathcal{U}_n 在這一步驟被消掉，成為 condition. 代表後面的推導也跟講義的 soft-margin SVM dual problem 一樣

By Slide L11, P30.

$$\min_{\alpha} \frac{1}{2} \sum \sum \alpha_n \alpha_m y_n y_m z_n^T z_m - \sum_{n=1}^N \alpha_n \quad \# \text{ 改述 [a]}$$

P10, B

$$\begin{aligned}& \int_0^1 \left(\frac{1}{2}(1-\rho)^2 - (1-\rho) \right)^2 \cdot \rho \, d\rho \\&= \int_0^1 \left(\frac{1}{2}(\rho^2 - 2\rho + 1) - 1 + \rho \right)^2 \cdot \rho \, d\rho \\&= \int_0^1 \left(\frac{\rho^2 - 1}{2} \right)^2 \cdot \rho \, d\rho \\&= \int_0^1 \left(\frac{\rho^4 - 2\rho^2 + 1}{4} \right) \cdot \rho \, d\rho \\&= \int_0^1 \frac{\rho^5 - 2\rho^3 + \rho}{4} \\&= \frac{1}{4} \left[\frac{\rho^6}{6} - \frac{2\rho^4}{4} + \frac{\rho^2}{2} \right] \Big|_0^1 \\&= \frac{1}{4} \left(\frac{1}{6} - \frac{1}{2} + \frac{1}{2} \right) \\&= \frac{1}{24} \quad \text{故选 [B]}$$



PIL A

```
./svm-train -s 0 -t 0 -c 10 satimage_5vo.scale
```

```
# Transform data into '5' versus 'not 5'
#
s = ""

with open('libsvm/satimage.scale.t', 'r') as f:
    for line in f.readlines():
        if line[0] != '6':
            s += '1' + line[1:] # "not X"
        else:
            s += '2' + line[1:] # "is X"

with open('libsvm/satimage_6vo.scale.t', 'w') as f:
    f.write(s)
```

```
from collections import defaultdict
import math

ans = defaultdict(int)
sum_alpha = 0

with open("satimage_5vo.scale.model", 'r') as f:
    for line in f.readlines():
        if line.find(':') != -1: # It's a SV
            l_list = line.split("\n")[0].rstrip().split(" ")

            y_alpha = float(l_list[0])
            sum_alpha += abs(y_alpha)
            for element in l_list[1:]:
                idx, value = element.split(":")
                ans[int(idx)] += float(value)*y_alpha

# Get |w|
l_acc = 0
for i in ans:
    l_acc += ans[i]**2
w_len = math.sqrt(l_acc)
print(f"\n|w| = {w_len}\n")
```

P12.C

```
echo "===== 2 vs. not 2 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_2vo.scale
./svm-predict satimage_2vo.scale satimage_2vo.scale.model output.txt
echo "===== 3 vs. not 3 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_3vo.scale
./svm-predict satimage_3vo.scale satimage_3vo.scale.model output.txt
echo "===== 4 vs. not 4 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_4vo.scale
./svm-predict satimage_4vo.scale satimage_4vo.scale.model output.txt
echo "===== 5 vs. not 5 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_5vo.scale
./svm-predict satimage_5vo.scale satimage_5vo.scale.model output.txt
echo "===== 6 vs. not 6 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_6vo.scale
./svm-predict satimage_6vo.scale satimage_6vo.scale.model output.txt
```

PL3.E

```
echo "===== 2 vs. not 2 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_2vo.scale
./svm-predict satimage_2vo.scale satimage_2vo.scale.model output.txt
echo "===== 3 vs. not 3 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_3vo.scale
./svm-predict satimage_3vo.scale satimage_3vo.scale.model output.txt
echo "===== 4 vs. not 4 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_4vo.scale
./svm-predict satimage_4vo.scale satimage_4vo.scale.model output.txt
echo "===== 5 vs. not 5 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_5vo.scale
./svm-predict satimage_5vo.scale satimage_5vo.scale.model output.txt
echo "===== 6 vs. not 6 ====="
./svm-train -s 0 -t 1 -d 3 -c 10 -r 1 -g 1 satimage_6vo.scale
./svm-predict satimage_6vo.scale satimage_6vo.scale.model output.txt
```

P14,D

```
echo "===== C = 0.01 ====="
./svm-train -s 0 -t 2 -g 10 -c 0.01 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== C = 0.1 ====="
./svm-train -s 0 -t 2 -g 10 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== C = 1 ====="
./svm-train -s 0 -t 2 -g 10 -c 1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== C = 10 ====="
./svm-train -s 0 -t 2 -g 10 -c 10 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== C = 100 ====="
./svm-train -s 0 -t 2 -g 10 -c 100 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
```

P15.B

```
echo "===== r = 0.1 ====="
./svm-train -s 0 -t 2 -g 0.1 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== r = 1 ====="
./svm-train -s 0 -t 2 -g 1 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== r = 10 ====="
./svm-train -s 0 -t 2 -g 10 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== r = 100 ====="
./svm-train -s 0 -t 2 -g 100 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
echo "===== r = 1000 ====="
./svm-train -s 0 -t 2 -g 1000 -c 0.1 satimage_1vo.scale
./svm-predict satimage_1vo.scale.t satimage_1vo.scale.model output.txt
```

P16,A

```
# echo "===== r = 0.1 ====="
# ./svm-train -s 0 -t 2 -g 0.1 -c 0.1 -v 22.175 satimage_1vo.scale
# echo "===== r = 1 ====="
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
./svm-train -s 0 -t 2 -g 1 -c 0.1 -v 22.175 satimage_1vo.scale
# echo "===== r = 10 ====="
# ./svm-train -s 0 -t 2 -g 10 -c 0.1 -v 22.175 satimage_1vo.scale
# echo "===== r = 100 ====="
# ./svm-train -s 0 -t 2 -g 100 -c 0.1 -v 22.175 satimage_1vo.scale
# echo "===== r = 1000 ====="
# ./svm-train -s 0 -t 2 -g 1000 -c 0.1 -v 22.175 satimage_1vo.scale
```