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Solution(part a): Let α'_i and α'_j be the new values of α_i and α_j respectively. In order to satisfy the constraint $\sum_i \alpha_i y_i = 0$, we have to keep the sum of $\alpha_i y_i + \alpha_j y_j$ same as before. Which leads us to the following equation. Note that value of y_i can be either -1 or +1.

$$\begin{aligned}
\alpha'_i y_i + \alpha'_j y_j &= \alpha_i y_i + \alpha_j y_j \\
\alpha'_j y_j &= \alpha_i y_i + \alpha_j y_j - \alpha'_i y_i \\
\alpha'_j y_j y_j &= \alpha_i y_i y_j + \alpha_j y_j y_j - \alpha'_i y_i y_j \\
\alpha'_j &= \alpha_i y_i y_j + \alpha_j - \alpha'_i y_i y_j && \text{as } y_j y_j = 1 \\
\alpha'_j &= \alpha_j + y_i y_j (\alpha_i - \alpha'_i) \\
\alpha'_j &= \alpha_j + h(\alpha_i - \alpha'_i) && \text{where } h = y_i y_j
\end{aligned}$$

Let L be the lower bound and U be the upper bound for α'_i

Case(I): $h = 1$,

$\therefore \alpha'_i + \alpha'_j = k_1$, where k_1 is a constant.

- If $k_1 > C$, then $\max \alpha_i = C$ and $\min \alpha_i = k_1 - C$
- If $k_1 < C$, then $\min \alpha_i = k_1$ and $\max \alpha_i = 0$

$$\begin{aligned}
L &= \max(0, \alpha_i + \alpha_j - C) \\
U &= \min(C, \alpha_i + \alpha_j)
\end{aligned}$$

Case(II): $h = -1$,

$\therefore \alpha'_j - \alpha'_i = k_2$, where k_2 is a constant.

- If $k_2 > 0$, then $\max \alpha'_i = 0$, and $\min \alpha'_i = C - k_2$.
- If $k_2 < C$, then $\min \alpha'_i = -k_2$ and $\max \alpha'_i = C$

$$\begin{aligned}
L &= \max(0, \alpha_i - \alpha_j) \\
U &= \min(C, C + \alpha_i - \alpha_j)
\end{aligned}$$

Thus we now have to restrict α_i within its range i.e. $[L, U]$

- $\alpha'_i = L$, if $\alpha_i < L$
- $\alpha'_i = H$, if $\alpha_i > H$
- $\alpha'_i = \alpha'_i$ (The value which we will compute)

Solution(part b): Let f be the objective function for the dual formulation for support vector machine.

$$f = -\frac{1}{2}\alpha^T G\alpha + \alpha^T 1$$

In order to find the extremum (maximum or minimum) of the objective with respect to α_i , we find the derivative of our objective function with respect to α_i and equate it to zero.

$$\begin{aligned} \frac{df}{d\alpha_i} &= 0 \\ \frac{d}{d\alpha_i}(-\frac{1}{2}\alpha^T G\alpha + \alpha^T 1) &= 0 \\ \frac{d}{d\alpha_i}(-\frac{1}{2}\alpha^T G\alpha) + \frac{d}{d\alpha_i}\alpha^T 1 &= 0 \\ \frac{da}{d\alpha_i} \frac{d}{da}(-\frac{1}{2}\alpha^T G\alpha) + \frac{da}{d\alpha_i}\alpha^T 1 &= 0 \\ \frac{da}{d\alpha_i}(\frac{d}{da}(-\frac{1}{2}\alpha^T G\alpha)) + \frac{d}{d\alpha_i}(\alpha^T 1) &= 0 \\ [1 \quad -h] \left(\frac{d}{da}(-\frac{1}{2}(a^T H a + \tilde{a}^T q a + a^T q^T \tilde{a} + \tilde{a}^T \tilde{H} \tilde{a})) \right) + (1-h) &= 0 \\ [1 \quad -h] \left(-\frac{1}{2}(2H a + q^T \tilde{a} + q^T \tilde{a} + 0) \right) + (1-h) &= 0 \\ -[1 \quad -h] (H a + q^T \tilde{a}) + (1-h) &= 0 \quad \dots(\text{i}) \\ [1 \quad -h] (H a) &= (1-h) - [1-h](q^T \tilde{a}) \\ [1 \quad -h] \begin{bmatrix} G_{ii} & G_{ij} \\ G_{ji} & G_{jj} \end{bmatrix} [\alpha'_i & \alpha'_j] &= (1-h) - [1-h](q^T \tilde{a}) \\ [G_{ii} - hG_{ji} & G_{ij} - hG_{jj}] \begin{bmatrix} \alpha'_i \\ \alpha_j + s\alpha_i - s\alpha'_i \end{bmatrix} &= (1-h) - [1-h](q^T \tilde{a}) \\ G_{ii}\alpha'_i - hG_{ji}\alpha'_i + G_{ij}\alpha_j + hG_{ij}\alpha_i - sG_{ij}\alpha'_i - sG_{jj}\alpha_j - G_{jj}\alpha_i + G_{jj}\alpha'_i &= (1-h) - [1-h](q^T \tilde{a}) \\ \alpha'_i(G_{ii} - hG_{ji} - hG_{ij} + G_{jj}) + (\alpha_j + h\alpha_i)(G_{ij} - hG_{jj}) &= (1-h) - [1-h](q^T \tilde{a}) \quad \dots(\text{ii}) \\ \alpha'_i &= \frac{1-h - [1-h](q^T \tilde{a}) - (\alpha_j + h\alpha_i)(G_{ij} - hG_{jj})}{(G_{ii} - hG_{ji} - hG_{ij} + G_{jj})} \end{aligned}$$

Checking if its a maxima:

$$\begin{aligned} \frac{d^2 f}{d\alpha_i'^2} &< 0 \\ \frac{d^2}{d\alpha_i'^2}(-\frac{1}{2}\alpha^T G\alpha + \alpha^T 1) &< 0 \\ \frac{d}{d\alpha_i'}(-[1 \quad -h] (H a + q^T \tilde{a}) + (1-h)) &< 0 \quad \text{using (i)} \\ \frac{d}{d\alpha_i'} - (\alpha'_i(G_{ii} - hG_{ji} - hG_{ij} + G_{jj}) + (\alpha_j + h\alpha_i)(G_{ij} - hG_{jj})) + (1-h) - [1-h](q^T \tilde{a}) &< 0 \quad \text{using (ii)} \\ G_{ii} - hG_{ji} - hG_{ij} + G_{jj} &> 0 \end{aligned}$$

Solution(part c):

Algorithm 1 Local optimization General SVM

```
1: procedure UPDATE(i,j)
2:    $h \leftarrow y_i y_j$ ;
3:    $[L, U] \leftarrow \text{FETCHBOUNDS}()$ ;
4:   if CHECKIFMAXIMA() then                                      $\triangleright$  Using second derivative to check if its a maxima
5:      $\alpha'_i = \text{COMPUTEALPHA}()$ ;
6:     if  $\alpha'_i < L$  then
7:        $\alpha'_i \leftarrow L$ ;
8:     else
9:        $\alpha'_i \leftarrow U$ ;
10:  else
11:     $L_{objective} = \text{COMPUTEOBJECTIVE}(L)$ ;
12:     $U_{objective} = \text{COMPUTEOBJECTIVE}(U)$ ;
13:    if  $L_{objective} > U_{objective}$  then
14:       $\alpha'_i \leftarrow L$ ;
15:    else
16:       $\alpha'_i \leftarrow U$ ;
17:  if  $\text{abs}(\alpha_i - \alpha'_i) > 10^{-10}(\alpha_i + \alpha'_i + 10^{-10})$  then
18:     $\alpha'_j \leftarrow \alpha_j + h(\alpha_i - \alpha'_i)$ ;
19:     $\alpha = \text{UPDATEALPHA}(\alpha'_i)$ ;
20:     $b = \text{UPDATEINTERCEPT}()$ ;
21:     $nch \leftarrow 1$ ;
22:  else
23:     $nch \leftarrow 0$ ;
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