
Algorithm 1 Sequential Minimal Optimization SVM

procedure *smo_svm*(*input*[],*output*[],*tolerance*,*c*, ϵ)

INPUTS: *input* - SET OF N INPUT IMAGE VECTORS

output - SET OF N OUTPUT DECISIONS (-1 OR 1)

tolerance - CONVERGENCE TOLERANCE

c - SOFT MARGIN PARAMETER

ϵ - EPSILON INSENSITIVITY ZONE

$\alpha \leftarrow \emptyset$, *bias* $\leftarrow 0$, *error* $\leftarrow 0$

num_changed $\leftarrow 0$, *examine* $\leftarrow 1$

while (*num_changed* > 0 || *examine*) **do**

num_changed $\leftarrow 0$

if (*examine*) **then**

for (*i* = 0 \rightarrow *input.size*) **do**

num_changed += *examine_example*(*i*)

end for

else

for (*i* = 0 \rightarrow *input.size*) **do**

if ($\alpha[i] \neq 0 \ \&\& \ \alpha[i] \neq c$) **then**

num_changed += *examine_example*(*i*)

end if

end for

end if

if (*examine* == 1) **then**

examine $\leftarrow 0$

else if (*num_changed* == 0) **then**

num_changed $\leftarrow 1$

end if

end while

end procedure

Algorithm 2 `examine_example()`

procedure `examine_example`(i_2)**INPUTS:** i_2 - INDEX TO EXAMPLE IN TRAINING SET $p_2 \leftarrow \text{input}[i_2], y_2 \leftarrow \text{output}[i_2], \alpha_2 \leftarrow \alpha[i_2]$ **if** ($\alpha_2 > 0 \ \&\& \ \alpha_2 < c$) **then** $\text{compute_error}_2 \leftarrow \text{error}[i_2]$ **else** $\text{compute_error}_2 \leftarrow \text{compute_svm}(p_2) - y_2$ **end if** $r_2 \leftarrow \text{compute_error}_2 * y_2$

#Check Lagrangian multiplier outside of allowable bounds.

#This tests KKT conditions for 1st choice lagrangian

if ($(r_2 < -\text{tolerance} \ \&\& \ \alpha_2 < c) \ \&\& \ (r_2 > \text{tolerance} \ \&\& \ \alpha_2 > 0)$) **then**

#Choose 2nd Lagrangian multiplier to maximize size of step taken during joint optimization.

#Approximate step size by: $|E_1 - E_2|$ $i_1 \leftarrow 0, \text{max} \leftarrow 0$ **for** ($i = 0 \rightarrow \text{input.size}$) **do****if** ($\alpha[i] > 0 \ \&\& \ \alpha[i] < c$) **then** $\text{aux_error} \leftarrow \text{error}[i]$ $\text{step_size} \leftarrow |\text{error}[i] - \text{compute_error}_2|$ **if** ($\text{step_size} > \text{max}$) **then** $\text{max} \leftarrow \text{step_size}$ $i_1 \leftarrow i$ **end if****end if****end for****if** ($i_1 \geq 0 \ \&\& \ (\text{take_step}(i_1, i_2)) == \text{true}$) **then** $\text{return } 1$ **end if**

#Sometimes SMO cannot make positive progress with heuristic above. So, iterate through

#non-bound examples, searching for second example that makes positive progress

 $\text{rnd} \leftarrow \text{random}(\text{input.size})$ **for** ($i_1 = \text{rnd} \rightarrow \text{input.size}$) **do****if** ($\alpha[i_1] > 0 \ \&\& \ \alpha[i_1] < c$) **then****if** ($(\text{take_step}(i_1, i_2)) == \text{true}$) **then** $\text{return } 1$ **end if****end if****end for****for** ($i_1 = 0 \rightarrow \text{rnd}$) **do****if** ($\alpha[i_1] > 0 \ \&\& \ \alpha[i_1] < c$) **then****if** ($(\text{take_step}(i_1, i_2)) == \text{true}$) **then** $\text{return } 1$ **end if****end if****end for**

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#If none of the non-bound examples lead to positive progress, SMO iterates through entire
#training set until positive progress. If this still doesn't work, abandon this example and move
#to next iteration for a better one
rnd  $\leftarrow$  random(input.size)
for (i1 = rnd  $\rightarrow$  input.size) do
    if (take_step(i1, i2)) == true then
        return 1
    end if
end for
for (i1 = 0  $\rightarrow$  rnd) do
    if (take_step(i1, i2)) == true then
        return 1
    end if
end for
else
    return 0
end if
end procedure
```

Algorithm 3 take_step()

```
procedure take_step( $i_1, i_2$ )  
  INPUTS:  $i_1$  - 1ST INDEX TO EXAMPLE IN TRAINING SET  
            $i_2$  - 2ND INDEX TO EXAMPLE IN TRAINING SET  
  if ( $i_1 == i_2$ ) then  
    return false  
  end if  
   $p_1 \leftarrow \text{input}[i_1], y_1 \leftarrow \text{output}[i_1], \alpha_1 \leftarrow \alpha[i_1]$   
  if ( $\alpha_1 > 0 \ \&\& \ \alpha_1 < c$ ) then  
     $\text{compute\_error}_1 \leftarrow \text{error}[i_1]$   
  else  
     $\text{compute\_error}_1 \leftarrow \text{compute\_svm}(p_1) - y_1$   
  end if  
   $p_2 \leftarrow \text{input}[i_2], y_2 \leftarrow \text{output}[i_2], \alpha_2 \leftarrow \alpha[i_2]$   
  if ( $\alpha_2 > 0 \ \&\& \ \alpha_2 < c$ ) then  
     $\text{compute\_error}_2 \leftarrow \text{error}[i_2]$   
  else  
     $\text{compute\_error}_2 \leftarrow \text{compute\_svm}(p_2) - y_2$   
  end if  
   $s \leftarrow y_1 * y_2$   
  if ( $y_1 \neq y_2$ ) then  
     $L \leftarrow \max(0, \alpha_2 - \alpha_1)$   
     $H \leftarrow \min(c, c + \alpha_2 - \alpha_1)$   
  else  
     $L \leftarrow \max(0, \alpha_2 + \alpha_1 - c)$   
     $H \leftarrow \min(c, \alpha_2 + \alpha_1)$   
  end if  
  if ( $L == H$ ) then  
    return false  
  end if  
   $k_{11} \leftarrow \text{gaussian\_kernel}(p_1, p_1)$   
   $k_{12} \leftarrow \text{gaussian\_kernel}(p_1, p_2)$   
   $k_{22} \leftarrow \text{gaussian\_kernel}(p_2, p_2)$   
   $\eta \leftarrow k_{11} + k_{22} - 2 * k_{12}$   
  if ( $\eta > 0$ ) then  
     $a_2 \leftarrow \alpha_2 + y_2 * \frac{\text{compute\_error}_1 - \text{compute\_error}_2}{\eta}$   
    if ( $a_2 < L$ ) then  
       $a_2 \leftarrow L$   
    end if  
    if ( $a_2 > H$ ) then  
       $a_2 \leftarrow H$   
    end if  
  else  
     $f_1 \leftarrow y_1 * (\text{compute\_error}_1 + \text{bias}) - \alpha_1 * k_{11} - s * \alpha_2 * k_{12}$   
     $f_2 \leftarrow y_2 * (\text{compute\_error}_2 + \text{bias}) - s * \alpha_1 * k_{12} - \alpha_2 * k_{22}$   
     $L_1 \leftarrow \alpha_1 + s * (\alpha_2 - L)$   
     $H_1 \leftarrow \alpha_1 + s * (\alpha_2 - H)$   
     $\Psi_L \leftarrow L_1 * f_1 + L * f_2 + \frac{1}{2} * L_1^2 * k_{11} + \frac{1}{2} * L^2 * k_{22} + s * L * L_1 * k_{12}$   
     $\Psi_H \leftarrow H_1 * f_1 + H * f_2 + \frac{1}{2} * H_1^2 * k_{11} + \frac{1}{2} * H^2 * k_{22} + s * H * H_1 * k_{12}$ 
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    if ( $\Psi_L < \Psi_H - \epsilon$ ) then
         $a_2 \leftarrow L$ 
    else if ( $\Psi_L > \Psi_H + \epsilon$ ) then
         $a_2 \leftarrow H$ 
    else
         $a_2 \leftarrow \alpha_2$ 
    end if
end if
if ( $|a_2 - \alpha_2| < \epsilon * (a_2 + \alpha_2 + \epsilon)$ ) then
    return false
end if
 $a_1 \leftarrow \alpha_1 + s * (\alpha_2 - a_2)$ 
if ( $a_1 < 0$ ) then
     $a_2 += s * a_1$ 
     $a_1 \leftarrow 0$ 
else if ( $a_1 > c$ ) then
     $a_2 += s * (a_1 - c)$ 
     $a_1 \leftarrow c$ 
end if
#Update bias
new_bias  $\leftarrow 0$ , delta_bias  $\leftarrow 0$ 
if ( $a_1 > 0 \ \&\& \ a_1 < c$ ) then
    new_bias  $\leftarrow compute\_error_1 + y_1 * (a_1 - \alpha_1) * k_{12} + y_2 * (a_2 - \alpha_2) * k_{22} + bias$ 
else
    #Both new Lagrangians at bound, choose threshold halfway between them
     $b_1 \leftarrow compute\_error_1 + y_1 * (a_1 - \alpha_1) * k_{11} + y_2 * (a_2 - \alpha_2) * k_{12} + bias$ 
     $b_2 \leftarrow compute\_error_2 + y_1 * (a_1 - \alpha_1) * k_{12} + y_2 * (a_2 - \alpha_2) * k_{22} + bias$ 
    new_bias  $\leftarrow \frac{b_1 + b_2}{2}$ 
end if
delta_bias  $\leftarrow new\_bias - bias$ 
bias  $\leftarrow new\_bias$ 
#Update error cache
for ( $i = 0 \rightarrow input.size$ ) do
    if ( $0 < \alpha[i] \ \&\& \ \alpha[i] < c$ ) then
        error[i]  $+= y_1 * (a_1 - \alpha_1) * gaussian\_kernel(p_1, input[i])$ 
         $+ y_2 * (a_2 - \alpha_2) * gaussian\_kernel(p_2, input[i])$ 
         $- delta\_bias$ 
    end if
end for
error[i1]  $\leftarrow 0$ 
error[i2]  $\leftarrow 0$ 
#Update Lagrangians
 $\alpha[i_1] \leftarrow a_1$ 
 $\alpha[i_2] \leftarrow a_2$ 
return true
end procedure
```

Algorithm 4 `compute_svm()`

```
procedure compute_svm( $p_1$ )  
  INPUTS:  $p_1$  - POINT TO COMPUTE DISTANCE AGAINST  
     $sum \leftarrow -bias$   
    for ( $i = 0 \rightarrow input.size$ ) do  
      if ( $\alpha[i] > 0$ ) then  
         $sum+ = \alpha[i] * output[i] * gaussian\_kernel(input[i], p_1)$   
      end if  
    end for  
    return sum  
end procedure
```

Algorithm 5 `gaussian_kernel()`

```
procedure gaussian_kernel( $input[i], p_1$ )  
  INPUTS:  $input[i]$  - 1ST POINT  
     $i_1$  - 2ND POINT  
     $gaussian \leftarrow \exp(-\frac{(|p_1 - p_2|)^2}{2 * \sigma^2})$   
    return gaussian  
end procedure
```
