



[Unit 2 Nonlinear Classification,](#)
[Linear regression, Collaborative](#)
[Course](#) > [Filtering \(2 weeks\)](#)
5. The Kernel Perceptron Algorithm

> [Lecture 6. Nonlinear Classification](#) >

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5. The Kernel Perceptron Algorithm

Computational Efficiency



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How the Kernel Perceptron Algorithm Works: Initialization

1/1 point (graded)

Recall that the original Perceptron Algorithm is given as the following:

Perceptron $\left(\{ (x^{(i)}, y^{(i)}), i = 1, \dots, n \}, T \right) :$

initialize $\theta = 0$ (vector);

for $t = 1, \dots, T,$

for $i = 1, \dots, n,$

if $y^{(i)} (\theta \cdot x^{(i)}) \leq 0,$

then update $\theta = \theta + y^{(i)} x^{(i)}.$

In the lecture, it was introduced that we can always express θ as

$$\theta = \sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)})$$

where values of $\alpha_1, \dots, \alpha_n$ may vary at each step of the algorithm. In other words, we can reformulate the algorithm so that we somehow initialize and update α_j 's, instead of θ .

The reformulated algorithm, or **kernel perceptron**, can be given in the following form:

Kernel Perceptron $\left(\{ (x^{(i)}, y^{(i)}), i = 1, \dots, n, T \} \right)$

Initialize $\alpha_1, \alpha_2, \dots, \alpha_n$ to some values;

for $t = 1, \dots, T$

for $i = 1, \dots, n$

if (Mistake Condition Expressed in α_j)

Update α_j appropriately

Look at the initialization statement of the algorithm. Which of the following is an equivalent way to initialize $\alpha_1, \alpha_2, \dots, \alpha_n$, if we want the same result as initializing $\theta = 0$?

☐ $\alpha_1 = \dots = \alpha_n = \theta$

☐ $\alpha_1 = \dots = \alpha_n = 1$

☒ $\alpha_1 = \dots = \alpha_n = 0$

☐ $\alpha_1 = \dots = \alpha_n = -1$



Solution:

Since $\theta = \sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)})$, setting $\alpha_j = 0$ for all j leads to $\theta = 0$.

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You have used 1 of 1 attempt

i Answers are displayed within the problem

How the Kernel Perceptron Algorithm Works: The Update

1/1 point (graded)

As in the previous problem, our goal is to correctly reformulate the original perceptron algorithm. In other words, we want the algorithm to be about updating α_j 's instead of θ .

Kernel Perceptron $\left(\left\{(x^{(i)}, y^{(i)}), i = 1, \dots, n, T\right\}\right)$

```

initialize  $\alpha_1, \alpha_2, \dots, \alpha_n$  to some values;
for  $t = 1, \dots, T$ 
  for  $i = 1, \dots, n$ 
    if (Mistake Condition Expressed in  $\alpha_j$ )
      Update  $\alpha_j$  appropriately

```

Now look at the line "**Update α_j appropriately**" in the above algorithm. Remember that we express θ as

$$\theta = \sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)})$$

Assuming that there was a mistake in classifying the i th data point i.e.

$$y^{(i)} (\theta \cdot x^{(i)}) \leq 0$$

which of the following conditions about $\alpha_1, \dots, \alpha_n$ is equivalent to

$$\theta = \theta + y^{(i)} \phi(x^{(i)}),$$

the update condition of the original algorithm?

☒ $\alpha_i = \alpha_i + 1$

☐ $\alpha_i = \alpha_i - 1$

☐ $\alpha_j = \alpha_j + 1$ for all $j \in 1, \dots, n$



Solution:

Expand θ in the last equation and it turns out only α_i gets updated:

$$\alpha_i y^{(i)} \phi(x^{(i)}) + y^{(i)} \phi(x^{(i)}) = (\alpha_i + 1) y^{(i)} \phi(x^{(i)}).$$

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How the Kernel Perceptron Algorithm Works: The Mistake Condition

1/1 point (graded)

Kernel Perceptron $\left(\{ (x^{(i)}, y^{(i)}), i = 1, \dots, n, T \} \right)$

initialize $\alpha_1, \alpha_2, \dots, \alpha_n$ to some values;

for $t = 1, \dots, T$

for $i = 1, \dots, n$

if (Mistake Condition Expressed in α_j)

Update α_j appropriately

Now look at the line "**Mistake Condition Expressed in α_j** " in the above algorithm. Remember that we express θ as

$$\theta = \sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)})$$

Which of the following conditions is equivalent to $y^{(i)} (\theta \cdot \phi(x^{(i)})) \leq 0$?

Remember from the video lecture above that given feature vectors $\phi(x)$ and $\phi(x')$, we define the Kernel function K as

$$K(x, x') = \phi(x) \phi(x').$$

☒ $y^{(i)} \sum_{j=1}^n \alpha_j y^{(j)} K(x^j, x^i) \leq 0$

☐ $y^{(i)} \sum_{j=1}^n \alpha_i y^{(j)} K(x^j, x^i) \leq 0$

☐ $y^{(i)} \sum_{j=1}^n \alpha_j y^{(i)} K(x^j, x^i) \leq 0$

☐ $y^{(i)} \sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)}) \leq 0$



Solution:

Substitute θ with $\sum_{j=1}^n \alpha_j y^{(j)} \phi(x^{(j)})$ in $y^{(i)} (\theta \cdot \phi(x^{(i)})) \leq 0$.

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You have used 1 of 1 attempt

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? Doesn't this make each Perceptron epoch $O(n^2)$

5

💬	6.5 Segment notes	6
👤	Community TA	
💬	Kernel	6
	Not very clear about what kernels are doing and why do we need to multiply by the number ...	
💬	Under the Kernel method what are we actually returning as solution?	3
	Under the simple perceptron, at the end the perceptron returns a Theta vector with all the co...	
?	Staff question (1)	2
	Can you review the solution for this question 1? **If Theta init =0** then in the first step of lo...	
?	[staff]	2
?	Concept of Similarity	5
	Prof. Jaakkola briefly discusses the kernel function as a sort of similarity measure. Maybe the ...	

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