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6. The Realizable Case - Quadratic program

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6. The Realizable Case - Quadratic program

The Realizable Case - Quadratic program



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The realizable case 1

1/1 point (graded)

In the realizable case, which of the following is true?

- ☐ There is exactly one (θ, θ_0) that satisfies $y^{(i)} (\theta \cdot x^{(i)} + \theta_0) \geq 1$ for $i = 1, \dots, n$.
- ☐ There are more than one, but finite number of (θ, θ_0) that satisfy $y^{(i)} (\theta \cdot x^{(i)} + \theta_0) \geq 1$ for $i = 1, \dots, n$.
- ☒ There are infinitely many (θ, θ_0) that satisfy $y^{(i)} (\theta \cdot x^{(i)} + \theta_0) \geq 1$ for $i = 1, \dots, n$.

**Solution:**

Without any additional constraint, because θ and θ_0 are continuous, there are numerous many (θ, θ_0) that satisfy the zero-error case.

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You have used 1 of 2 attempts

i Answers are displayed within the problem

The realizable case 2

1/1 point (graded)

Remember the objective function

$$J(\theta, \theta_0) = \frac{1}{n} \sum_{i=1}^n \text{Loss}_h(y^{(i)}(\theta \cdot x^{(i)} + \theta_0)) + \frac{\lambda}{2} \|\theta\|^2$$

In the realizable case, we can always find (θ, θ_0) such that the sum of the hinge losses is 0. In this case, what does the objective function J reduce to?

☐ $\frac{1}{n} \sum_{i=1}^n \text{Loss}_h(y^{(i)}(\theta \cdot x^{(i)} + \theta_0))$

☐ $\frac{1}{n} \sum_{i=1}^n \text{Loss}_h(y^{(i)}(\theta \cdot x^{(i)} + \theta_0)) + \frac{\lambda}{2} \|\theta\|^2$

☒ $\frac{1}{2} \|\theta\|^2$



Solution:

In the realizable case, we can always find a decision boundary such that the first term of $J(\theta, \theta_0)$ is 0. Thus $J(\theta, \theta_0)$ reduces to $\frac{\lambda}{2} \|\theta\|^2$. Our goal is to find θ that minimizes J anyways, so J reduces to $\frac{1}{2} \|\theta\|^2$

You have used 1 of 2 attempts

i Answers are displayed within the problem

Support Vectors

1/1 point (graded)

Support vectors refer to points that are exactly on the margin boundary. Which of the following is true? Choose all those apply.

☐ If we remove one point that is not a support vector, we will get a different θ, θ_0

☒ If we remove all points that are support vectors, we will get a different θ, θ_0

☐ If we remove one point that is a support vector, we will get the same θ, θ_0

☒ If we remove one point that is not a support vector, we will get the same θ, θ_0



Solution:

Support vectors determine the exact solution θ, θ_0 that minimizes $J(\theta, \theta_0)$. Thus removing/changing all of them changes the θ, θ_0 . On the other hand, any training example that is not a support vector has no influence on θ, θ_0 . Thus removing/changing them does not affect θ, θ_0 .

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You have used 1 of 2 attempts

i Answers are displayed within the problem

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☒ Not sure about the answer of "Realisable case 2"

4

Community TA

✓	Quadratic programming	3
	As part of this course are we going to implement any kind of the quadratic program? I have n...	
?	[staff]: Support Vectors problem	1 new_ 3
💬	what is "realizable case"?	3
	You've never defined the term "realizable case". In fact, the lecture includes this sentence: "In...	
💬	The "simple" quadratic case exposed is equivalent to the full minimisation of the $J(\theta)$ function with very small λ	1
	Community TA	
💬	Question 4	3
	Third option of question 4 may be a little confusing without additional information. What hap...	
?	About the last question about exercise 4	2
	My question is, if you mean "removing support vector", it means that there are more than 1 s...	
?	Are Linear Support Vector Machines good only for linearly separable sets of data?	2
	Since we constrain the SVM to hard margins without hinge loss does that mean we do not all...	
💬	[staff]: Support Vectors - I don't agree with the correct answer.	2
?	Points that are not support vectors	2
	Hi, Is it true to say that the points that are not support vectors will only affect the convergenc...	
💬	Lecture never shows us how to solve theta	1

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