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### **Machine Learning with Python-From Linear Models to Deep Learning**

Discussion Course **Progress** Dates Resources

☆ Course / Unit 2. Nonlinear Classification, Linear regression, Col... / Lecture 5.



## 8. Regularization

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Exercises due Mar 8, 2023 08:59 -03 Completed

### **Ridge Regression**



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#### Video

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#### **Transcripts**

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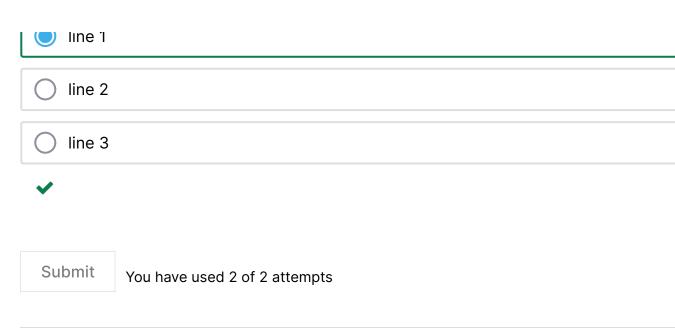
## Regularization: extreme case 1

1/1 point (graded)

As in the video above, define the loss function

$$J_{n,\lambda}\left( heta, heta_0
ight) = rac{1}{n}\sum_{t=1}^nrac{\left(y^{(t)}- heta\cdot x^{(t)}- heta_0
ight)^2}{2} + rac{\lambda}{2}\| heta\|^2$$

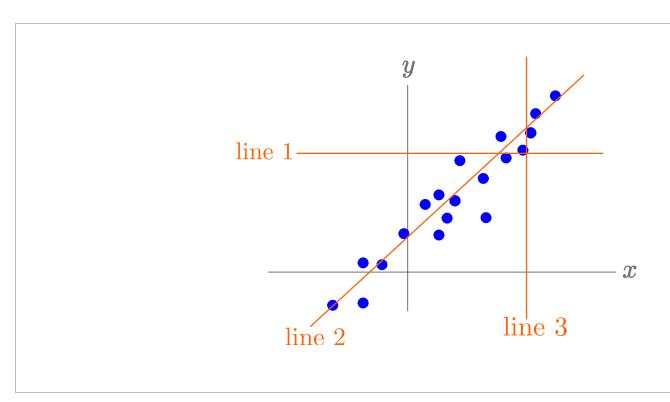
where  $\pmb{\lambda}$  is the regularization factor.



## Regularization: Extreme case 2

1/1 point (graded)
As in the problem above,

where is the regularization factor.



In the figure above, the blue dots are the training examples. If we decrease to , so (shorthand for the weights optimal with respect to and the depicted data) also changed does the predictor line (i.e., the graph of the function from s to s

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Regularization with large data sets

Let us assume we have a very large data set at hand, and a not-so-complex model. Do we still need to do re

✓ possible typo

In part two, should the above question say "if we decrease lambda to negative infinity..."? I can't really make

- taking derivative of theta while it is a norm
- Found an error in the transcript

\*\*ridge\*\* regression not \*reach\* regression. 0:17:)

? mistake information in "Regularization: Extreme case 2"

Regularization: Extreme case 2: "In the figure above, the blue dots are the training examples. If we decrease

possible plotting issue!

Shouldn't Line 1 have been drawn lower, so that its location corresponds to the average y values?

Need intuition on the regularization term

Why is the regularization term defined in terms of norm(theta)? For linear classifiers, it makes sense why reg

8:38 (1-eta\*lambda)\*theta expression

I don't understand why "at every point of our average we're actually pushing thetas down" if the learning rate

? If regularization is high, the points are not adjusted as much as if regularization is low, the point adjusted a lot?

So how does this translate to the graph as shown above? Like if the points get adjusted, how will i know if it

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