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Machine Learning with Python-From Linear Models to Deep Learning

<u>Help</u>



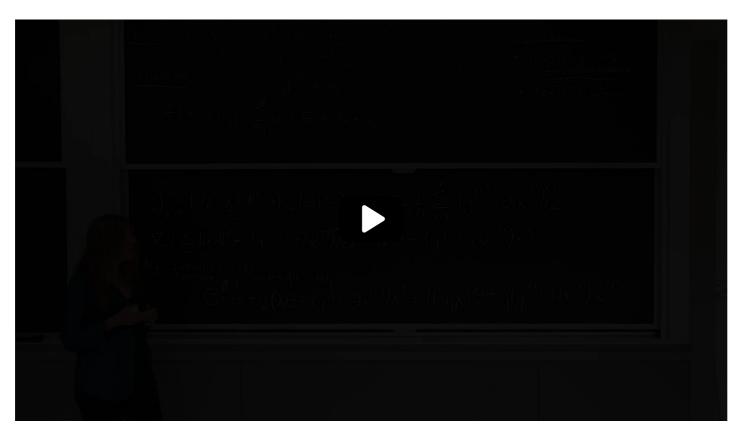
lala sci <u>sandipan dey</u>

Unit 2 Nonlinear Classification, Linear regression, Collaborative

<u>Course</u> > <u>Filtering (2 weeks)</u>

> <u>Lecture 5. Linear Regression</u> > 8. Regularization

# 8. Regularization **Ridge Regression**



this new reach

regression formula, apply gradient-based algorithm,

and to find the best theta.

In the homework, you will see-- in the exercise, you will see how you can do the same kind of modification,

very straightforward modification,

to get a solution for this objective for the closed form

algorithm.

X 9:51 / 9:51 CC ▶ Speed 1.50x

End of transcript. Skip to the start.

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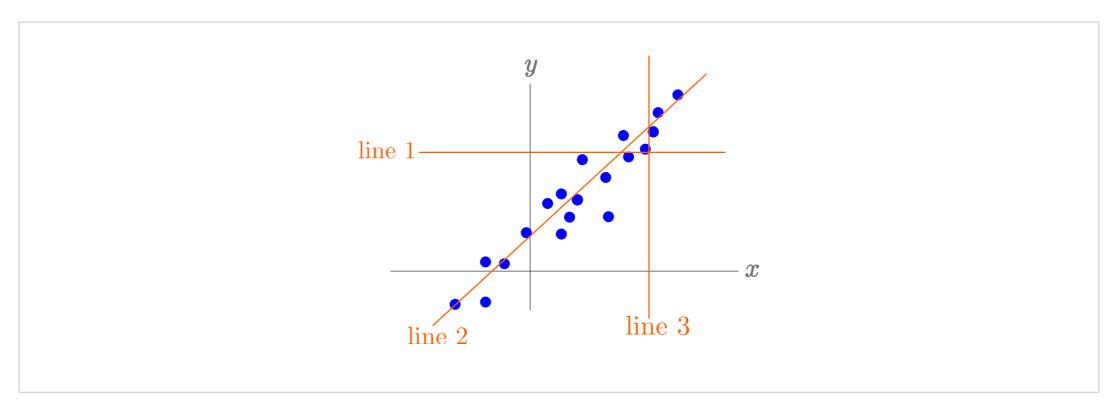
## Regularization: extreme case 1

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  $\lambda$  is the regularization factor.



In the figure above, the blue dots are the training examples. If we increase  $\lambda$  to  $\infty$ , where does  $f(x)= heta\cdot x+ heta_0$  converge to?

● line 1

O line 2

line 3

#### **Solution:**

If we increase  $\lambda$  to  $\infty$ , minimizing J is equivalent to minimizing  $||\theta||$ . Thus  $\theta$  will have to be a zero vector. Thus  $f(x) = \theta \cdot x + \theta_0$  becomes  $f(x) = \theta_0$ , a horizontal line. Thus f converges to line 1.

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You have used 2 of 2 attempts

**1** Answers are displayed within the problem

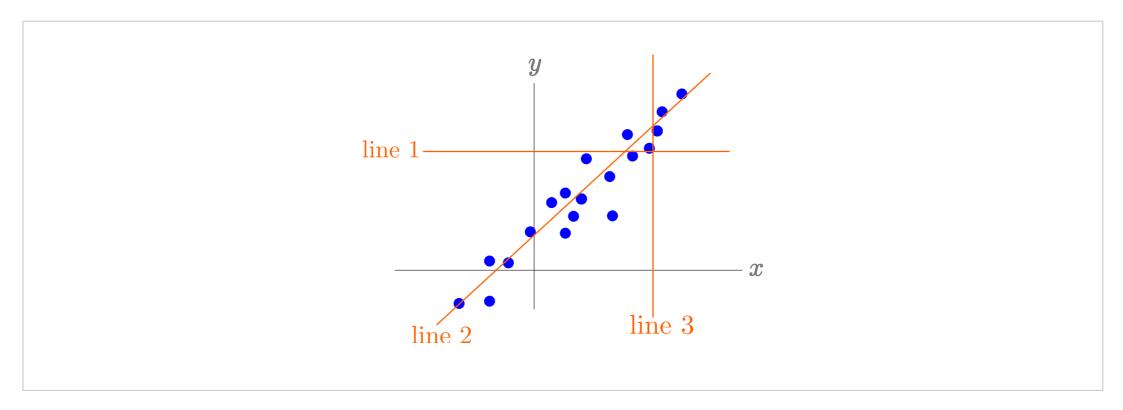
### Regularization: Extreme case 2

1/1 point (graded)

As in the problem above,

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

where  $\lambda$  is the regularization factor.



In the figure above, the blue dots are the training examples. If we decrease  $\lambda$  to 0, where does  $f(x)= heta\cdot x+ heta_0$  converge to?

- O line 1
- line 2 ✓
- O line 3

### **Solution:**

If we decrease  $\lambda$  to zero, minimizing J is equivalent to minimizing  $\frac{1}{n}\sum_{t=1}^n\frac{(y^{(t)}-\theta\cdot x^{(t)}-\theta_0)^2}{2}$ , which is the "fit." Thus f converges to line 2.

Submit

You have used 1 of 2 attempts

**1** Answers are displayed within the problem

Discussion

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**Topic:** Unit 2 Nonlinear Classification, Linear regression, Collaborative Filtering (2 weeks):Lecture 5. Linear Regression / 8. Regularization

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? Why the term "regularization"? What is "regular" about the modified objective function? Is there some deeper reason why the term "regularization" is used? For example how we distinguish perturbations as	1
Maybe a typo 00:00:17,320> 00:00:20,950 And it's called 【reach regression】. br/> 138 138 138 138 138 130:00:20,110> 00:09:27,340 100:00:27,340 100	1
? Why adding regularization can help the model become more general?  I notice that adding regularization can make the parameter more close to zero. How I still confused about why the parameter more close to zero, the more general the model	2
? Regularization: Extreme case 2: the answer explanation is missing theta 0	3
Regularization: theta_0 left out of regularization, is this typical?	3
Regulatization only effects positive theta updates to prevent increasing theta values but not decreasing of theta value?  Can someone please clarify, if regularization effects only when theta updates are positive and not when theta update is negative, that is it does not effect update step when t	4

<u>Staff] Regularization: extreme case 1 and 2</u>

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