

<u>Unit 2 Nonlinear Classification</u>, <u>Linear regression, Collaborative</u>

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

5. Gradient Based Approach

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

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5. Gradient Based Approach Learning Algorithm: Gradient Based Approach





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True or False

0 points possible (ungraded)

Let $R_{n}\left(heta
ight)$ be the least squares criterion defined by

$$R_{n}\left(heta
ight) = rac{1}{n}\sum_{t=1}^{n}\operatorname{Loss}\left(y^{(t)} - heta \cdot x^{(t)}
ight).$$

Which of the following is true? Choose all those apply.

- $lacksquaremath{oldsymbol{arphi}}$ The least squares criterion $R_n\left(heta
 ight)$ is a sum of functions, one per data point.
- Stochastic gradient descent is slower than gradient descent.
- $lacksquare
 abla_{ heta} R_n\left(heta
 ight)$ is a sum of functions, one per data point.



Solution:

For every point, the loss is a function of θ , so the least squares criterion $R_n\left(\theta\right)$ is a sum of functions, one per data point, and this is what makes stochastic gradient descent possible. We want to do stochastic gradient descent because it is faster than gradient descent. Finally, because $R_n\left(\theta\right)$ is sum of functions, one per data point, $\nabla_{\theta}R_n\left(\theta\right)$ is also a sum of functions one per data point.

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You have used 1 of 3 attempts

1 Answers are displayed within the problem

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