

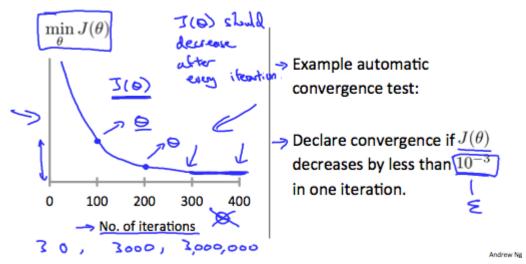
Gradient Descent in Practice II - Learning Rate

Note: [5:20 - the x -axis label in the right graph should be θ rather than No. of iterations]

Debugging gradient descent. Make a plot with *number of iterations* on the x-axis. Now plot the cost function, $J(\theta)$ over the number of iterations of gradient descent. If $J(\theta)$ ever increases, then you probably need to decrease α .

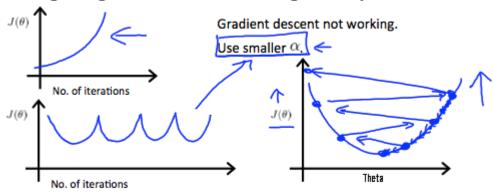
Automatic convergence test. Declare convergence if J(θ) decreases by less than E in one iteration, where E is some small value such as 10^{-3} . However in practice it's difficult to choose this threshold value.

Making sure gradient descent is working correctly.



It has been proven that if learning rate α is sufficiently small, then J(θ) will decrease on every iteration.

Making sure gradient descent is working correctly.



- For sufficiently small α , $J(\theta)$ should decrease on every iteration. \leq
- But if lpha is too small, gradient descent can be slow to converge.

To summarize:

If α is too small: slow convergence.

If α is too large: may not decrease on every iteration and thus may not converge.

イ 完成

