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imes Lessons

This Course: Machine Learning

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Gradient Checking

Gradient checking will assure that our backpropagation works as intended. We can approximate the derivative of our cost function with:

$$\frac{\partial}{\partial \Theta} J(\Theta) pprox \frac{J(\Theta + \epsilon) - J(\Theta - \epsilon)}{2\epsilon}$$

With multiple theta matrices, we can approximate the derivative **with respect to** Θ_i as follows:

$$rac{\partial}{\partial \Theta_{j}} rac{J(\Theta)}{2\epsilon} pprox rac{J(\Theta_{1}, \ldots, \Theta_{j} + \epsilon, \ldots, \Theta_{n}) - J(\Theta_{1}, \ldots, \Theta_{j} - \epsilon, \ldots, \Theta_{n})}{2\epsilon}$$

A small value for ϵ (epsilon) such as $\epsilon=10^{-4}$, guarantees that the math works out properly. If the value for ϵ is too small, we can end up with numerical problems.

Hence, we are only adding or subtracting epsilon to the Θ_j matrix. In octave we can do it as follows:

We previously saw how to calculate the deltaVector. So once we compute our gradApprox vector, we can check that gradApprox \approx deltaVector.

Once you have verified **once** that your backpropagation algorithm is correct, you don't need to compute gradApprox again. The code to compute gradApprox can be very slow.

✓ Complete



