coursera

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Cost Function and Backpropagation

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- Video: Backpropagation
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Backpropagation in Practice

- Video: Implementation
 Note: Unrolling Parameters
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Gradient Checking

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

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

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

$$egin{aligned} rac{\partial}{\partial \Theta_j} J(\Theta) pprox \ rac{J(\Theta_1, \dots, \Theta_j + \epsilon, \dots, \Theta_n) - J(\Theta_1, \dots, \Theta_j - \epsilon, \dots, \Theta_n}{2\epsilon} \end{aligned}$$

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:

```
1  epsilon = 1e-4;
2  for i = 1:n,
3    thetaPlus = theta;
4    thetaPlus(i) += epsilon;
5    thetaMinus = theta;
6    thetaMinus(i) -= epsilon;
7    gradApprox(i) = (J(thetaPlus) - J(thetaMinus))/(2*, end;
9
```

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.

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