

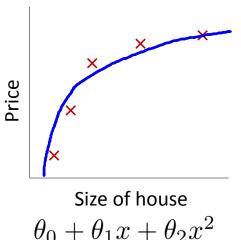
Regularization

Cost function

Machine Learning



problema? apesar de funçao corresponder nos pontos temos grandes variancias(+ minimos) ->



Size of house
$$\theta_0 + \theta_1 x + \theta_2 x^2 + \theta_3 x^3 - \theta_4 x^3 - \theta_5 x^$$

Suppose we penalize and make θ_3 , θ_4 really small.





so podemos minimizar esta funçao se theta3 e theta4 forem pequenos

Regularization.

Small values for parameters $\theta_0, \theta_1, \dots, \theta_n$

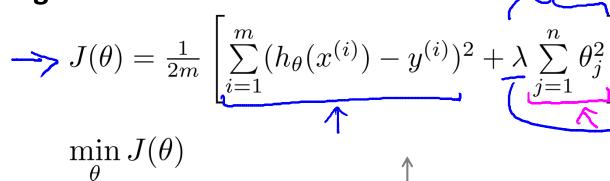
- "Simpler" hypothesis
- Less prone to overfitting <

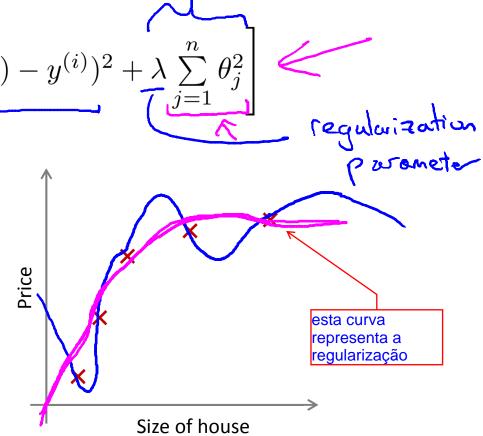
Housing:

- Features: x_1, x_2, \dots, x_{100}
- Parameters: $\theta_0, \theta_1, \theta_2, \dots, \theta_{100}$

$$J(\theta) = \frac{1}{2m} \left[\sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)})^2 + \lambda \right]$$

Regularization.





In regularized linear regression, we choose θ to minimize

$$J(\theta) = \frac{1}{2m} \left[\sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)})^2 + \lambda \sum_{j=1}^{n} \theta_j^2 \right]$$

What if λ is set to an extremely large value (perhaps for too large for our problem, say $\lambda=10^{10}$)?

- Algorithm works fine; setting λ to be very large can't hurt it
- Algortihm fails to eliminate overfitting.
- Algorithm results in underfitting. (Fails to fit even training data well).
- Gradient descent will fail to converge.

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