





Optional Lab - Regularized Cost and Gradient

Goals

In this lab, you will:

- extend the previous linear and logistic cost functions with a regularization term.
- rerun the previous example of over-fitting with a regularization term added.

```
In []: import numpy as np
%matplotlib widget
import matplotlib.pyplot as plt
from plt_overfit import overfit_example, output
from lab_utils_common import sigmoid
np.set_printoptions(precision=8)
```

Adding regularization

Regularized linear regression

$$\begin{aligned} & \underset{\overrightarrow{w},b}{\min} \ J(\overrightarrow{w},b) = \underset{\overrightarrow{w},b}{\min} \left(\frac{1}{2m} \sum_{i=1}^{m} (f_{\overrightarrow{w},b}(\overrightarrow{x}^{(i)}) - y^{(i)})^2 + \frac{\lambda}{2m} \sum_{j=1}^{n} w_j^2 \right) \\ & \text{Gradient descent} \\ & \text{Repeat } \left\{ \\ & w_j = w_j - \alpha \left(\frac{\partial}{\partial w_j} J(w_j,b)\right) \right. \\ & = \frac{1}{m} \sum_{i=1}^{m} \left(f_{\overrightarrow{w},b}(\overrightarrow{x}^{(i)}) - y^{(i)}\right) x_j^{(i)} \right. \\ & + \frac{\lambda}{m} \ \text{w}_j \\ & b = b - \alpha \left(\frac{\partial}{\partial b} J(w_j,b)\right) \\ & + \frac{1}{m} \sum_{i=1}^{m} \left(f_{\overrightarrow{w},b}(\overrightarrow{x}^{(i)}) - y^{(i)}\right) x_j^{(i)} \right. \\ & + \frac{\lambda}{m} \ \text{w}_j \\ & + \frac{\lambda}{m} \ \text{w}$$

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