

# Regularization and the Two Norms

Demo for the Beamer Template

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# 1 Definition

# What is Regularization

## Regularization

“Penalizing the weights and biases of the error function to keep them from reaching large values is called Regularization [1]”.

Mathematically,

$$\underset{x}{\text{minimize}} \quad \|Ax - b\|_2^2 + \lambda \|x\|_n \quad (1)$$

The last term of equation 1 is the regularization term here.

## 2 The Two Norms

# Two Norms of Interest

There are two most widely used norms that we are interested in:

- ▶ The  $\ell_1$ -norm
- ▶ The  $\ell_2$ -norm

# The $\ell_1$ -Norm

Replacing  $n$  of equation 1 with 1, i.e. using the absolute values will result in  $\ell_1$ -norm. So, the following would be  $\ell_1$  regularized cost functions:

$$\underset{x}{\text{minimize}} \quad \|Ax - b\|_2^2 + \lambda \|x\|_1 \quad (2)$$

# The $\ell_2$ -Norm

Replacing  $n$  of equation 1 with 2, i.e. using the squared values of the weights, will result in  $\ell_2$ -norm. So, the following would be  $\ell_2$  regularized cost functions:

$$\underset{x}{\text{minimize}} \quad \|Ax - b\|_2^2 + \lambda \|x\|_2 \quad (3)$$



## ② The Two Norms

### Comparison Between the Two Norms

# Comparison

	$\ell_1$ -Norm	$\ell_2$ -Norm
Sparsity (See Fig 1)	Yes	No
Robustness	Yes	No

Table 1: Differences between the two norms

# Sparsity of $\ell_1$ -Norm

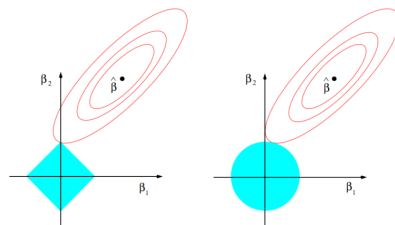


Figure 1: Contours of error and constraint functions for Lasso (Left) and Ridge Regression (Right) [2].

### 3 References

# References I

- [1] C. M. Bishop. *Pattern Recognition and Machine Learning (Information Science and Statistics)*. Berlin, Heidelberg: Springer-Verlag, 2006. ISBN: 0387310738.
- [2] J. Friedman, T. Hastie, R. Tibshirani, et al. *The elements of statistical learning*. Vol. 1. Springer series in statistics New York, 2001.

# Thank you!