

# Applying Regression and Resampling Techniques to Norwegian Terrain Data with Franke's Function as Test Function

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**Abstract**

## 1 Introduction

## 2 Theory

### 2.1 Datasets

We will analyze the following datasets:

#### 2.1.1 MNIST

The famous MNIST dataset is a collection of handwritten numbers, as  $28 \times 28$  grayscale images. It comes in two sets, a training set with 60,000 images, and a testing set with 10,000 images. In this report, we will model the inputs as a  $28 \times 28 = 784$ -dimensional vector, and the output as a 10-dimensional state vector, with each dimension representing the corresponding digit.

### 2.2 Stochastic Gradient Descent

*Gradient descent* describes the process of finding a local minimum of a function (the cost function, in our case) by following the negative value of the gradient at each point, stepwise. *Stochastic gradient descent* or SGD is a way of increasing the numerical efficiency of this process, by doing this process stochastically.

This involves randomly dividing the training data into a given number of *mini batches*. For each mini batch, the gradient is found by averaging the gradient

value each mini batch sample has. Then the weights and biases are updated (take a step down the "slope") and the process is repeated for the rest of the mini batches. The updating done at each mini batch is expressed mathematically as

$$w \rightarrow w' = w - \frac{\eta}{m} \sum_i^m \nabla C_{i,w}$$

$$b \rightarrow b' = b - \frac{\eta}{m} \sum_i^m \nabla C_{i,b},$$

where  $m$  is the number of datapoints in the mini batches and  $\nabla C_i$  is the gradient at each individual data point. After exhausting all the training data, we have finished a so-called *epoch*, of which we can perform as many as necessary.

### **3 Method**

### **4 Results**

### **5 Discussion**

### **6 Conclusion**

## A Appendix