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# Optimization Techniques Report

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## Abstract

Experimented on different optimization Techniques and findings are documented in this document.

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## 1 GRADIENT DESCENT

Gradient descent is a way to minimize an objective function  $J(\theta)$  parameterized by a model's parameters  $\theta \in \mathbb{R}^d$  by updating the parameters in the opposite direction of the gradient of the objective function  $\nabla_{\theta} J(\theta)$  with respect to the parameters. The learning rate  $\eta$  determines the size of the steps we take to reach a (local) minimum. In other words, we follow the direction of the slope of the surface created by the objective function downhill until we reach a valley. [1]

## 2 VARIATIONS OF GRADIENT DESCENT

- Stochastic (SGD)
- Mini-batch SGD
- Momentum
- Nesterov Accelerated (NAG)
- Adagrad
- Adadelata
- RMSprop
- Adam
- AdaMax
- Nadam
- AMSGrad
- AdamW
- Yogi
- RAdam
- Lookahead

## REFERENCES

- [1] Sebastian Ruder. "An overview of gradient descent optimization algorithms". In: *CoRR* abs/1609.04747 (2016). arXiv: 1609.04747. URL: <http://arxiv.org/abs/1609.04747>.