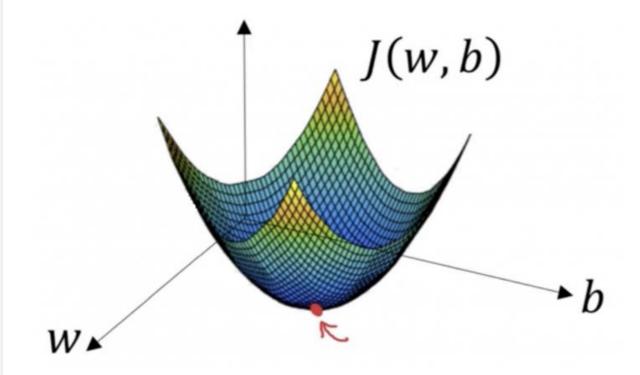
Machine Learning and Deep Learning Optimizers Implementation

Artificial Intelligence & Data Science

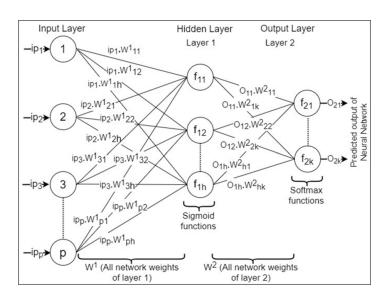
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• In ML&DL, the optimizers are the algorithms used to find the best values of model parameters that minimize the model prediction error (the minimum of loss function).



• Optimizers find the best model parameters of simple single variable and multivariable linear regression models as well as complex and deep learning models.

- Single Variable LR: $h_{\theta}(x) = \theta_{\theta} + \theta_{I} x$
- Multi Variable LR $h_{\theta}(x) = \theta_{\theta} x_{\theta} + \theta_{1} x_{1} + \theta_{2} x_{2} + \dots + \theta_{n} x_{n}$



- Almost all ML&DL optimizers are **numerical optimization** algorithms such as:
 - Batch Gradient Descent (GD).
 - Stochastic GD (SGD).
 - Mini-Batch GD.
 - Momentum based GD.
 - Nestrov Accelerated Gradient (NAG).
 - Adagrad.
 - RMSProp.
 - Adam.
 - BFGS.

• In order to use or even choose in the first place the suitable optimizer to train your model, adjusting hyperparameters, obtain fast and efficient training, and sometimes modify the optimization method, you must understand how the optimizer work.



Fully understanding of ML&DL optimizers is crucial.

It is your first and most important step before learning ML&DL.

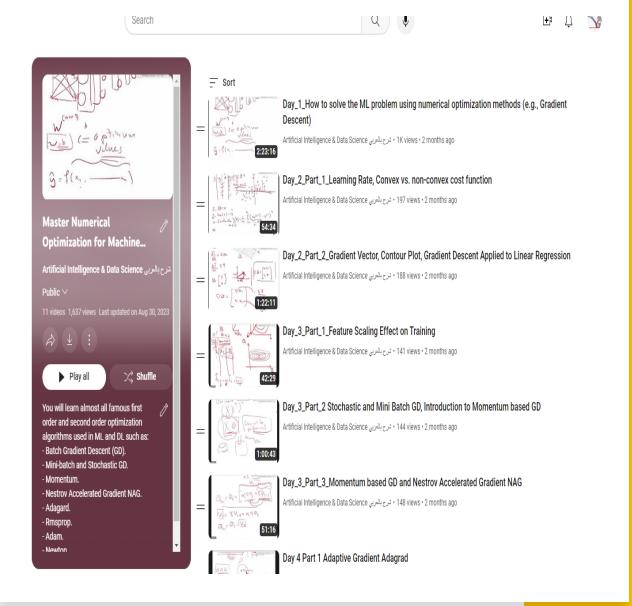
Master Numerical Optimization for ML&DL (NOFML&DL) in 5 Days

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Master NOFML&DL in 5 Days Course Link



Why Shall We Study Numerical Optimization? The Machine Learning Problem. Analytical, Algebraic and Numerical Solutions. **Numerical Optimization.** Single Variable Linear Regression. **Vector Norms and Loss Function** Gradient **Gradient Descent (GD) Algorithm.**

Convex Function.

Gradient of Multivariable Function.

Contour Plot.

Local vs. Global Minimum.

GD Applied to LR (Single Variable).

GD Applied to Multivariable LR.

Features Scaling.

Batch GD Problems.

GD Variants.

Stochastic GD.

Mini-Batch GD.

GD Challenges.

Advanced Optimizers.

Momentum-based GD.

Nesterov accelerated GD (NAG).

Adaptive Learning Rate.

Adagrad.

RMSProp.

Adam.

GD Connection to Taylor Series.

Second Order Optimization.

Newton's Method for Optimization.

Quasi-Newton Methods.

BFGS.

How to Understand the Optimizers?

- In order to understand the optimizers thoroughly, you need to implement it from scratch by yourself.
- See the effect of hyperparameters in training.
- Explore this effect by plotting the learning curves.



Machine Learning and Deep Learning Optimizers Implementation

PW 1: Implementation of Batch (Vanilla) GD for single variable LR model.

PW 2: Implementation of Batch (Vanilla) GD for Multi-variables LR model.

PW 3: Implementation of Stochastic GD for Single and Multi-variables LR model.

PW 4: Implementation of Mini-Batch GD for Single and Multi-variables LR model.

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PW 5: Implementation of Momentum based GD for Single and Multi-variables LR model.

PW 6: Implementation of NAG for Single and Multivariables LR model.

Machine Learning and Deep Learning Optimizers Implementation

PW 7: Implementation of Adagrad for Single and Multi-variables LR model.

PW 8: Implementation of RMSProp for Single and Multi-variables LR model.

PW 9: Implementation of Adam for Single and Multi-variables LR model.

PW 10: Implementation of BFGS method for Single and Multi-variables LR model.



Resources

Master numerical optimization for ML&DL in 5 Days course material.

Data files.

Guided steps code files.

Ideal solution code files.

Attention

- You must do the implementation by yourself.
- Do not jump directly to the ideal solutions.



