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by

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Abstract

Placeholder

Terms and Definitions

α	Learning Rate
η	Reduced Learning Rate
$\mathbf{I}$	Identity Matrix
$\odot$	Hadamard Product
NN	Neural Network
MSE	Mean Squared Error
MAE	Mean Absolute Error

Chapter 1

Introduction

Introduction

Chapter 2

Analysis of Bitcoin Price using Deep Learning Models

We ran a different experiment with the limited dataset containing only influencers but did not observe any major improvement from the previous experiment.

Chapter 3

Conclusion and Future Works

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