

# Collected Experiment Report: SimpleNN - DeepNN .

16/02/2025

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This is an automated report for the Experimen	nt with neural netv	works on traffic volume	e dataset; the following
models have been analyzed:			

○ Simp	leNN
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○ DeepNN

# **Experiment description:**

Experiment with neural networks on traffic volume dataset

#### **Model setup**

The models have been used for the following forecast purposes:

- one\_step
- multistep
- o recursive

The models have been optimized using the following hyperparameters:

- optimizer: ['adam', 'sgd']

- epochs: [100, 200]

- batch\_size: [32, 64]

- scaler: [None, StandardScaler(), MinMaxScaler(), RobustScaler(), PowerTransformer()]

And with the following search algorithms:

o grid

 $\quad \circ \, random \\$ 

The used performance measure is the neg\_mean\_absolute\_error measure.



#### **Dataset setup**

The baseline dataset used for these forecasts is

the 'Metro Interstate Traffic Volume with hourly features and holiday markings.' dataset: 'Metro Interstate Traffic Volume with hourly features and holiday markings.'.

The test size used for the forecasts is 0.2.

#### ODataset 1

- name: univariate\_lagged

- dataset\_type: univariate

prediction\_type: one\_step

- components: ['one\_step\_target', 'lagged\_target']

#### ODataset 2

- name: univariate\_temporal

- dataset\_type: univariate

prediction\_type: one\_step

- components: ['one\_step\_target', 'temporal\_features']

#### ODataset 3

name: multivariate\_lagged

- dataset\_type: multivariate

prediction\_type: one\_step

- components: ['one\_step\_target', 'lagged\_target', 'feature\_columns']

#### ODataset 4

- name: multivariate\_lagged\_temporal

- dataset\_type: multivariate

prediction\_type: one\_step

- components: ['one\_step\_target', 'temporal\_features', 'feature\_columns', 'lagged\_target']

#### ODataset 5



- name: univariate\_lagged\_multistep

- dataset\_type: univariate

- prediction\_type: multistep

- components: ['multistep\_target', 'lagged\_target']

#### ODataset 6

- name: multivariate\_lagged\_temporal\_multistep

- dataset\_type: multivariate

- prediction\_type: multistep

- components: ['multistep\_target', 'temporal\_features', 'feature\_columns', 'lagged\_target']

# Results

For the models; SimpleNN, DeepNN, the following models and datasets yielded the best results.



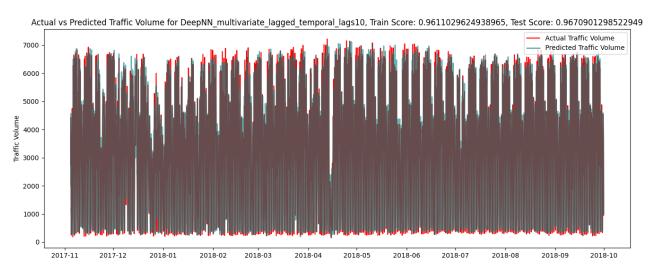
#### The best model for one\_step forecasting.

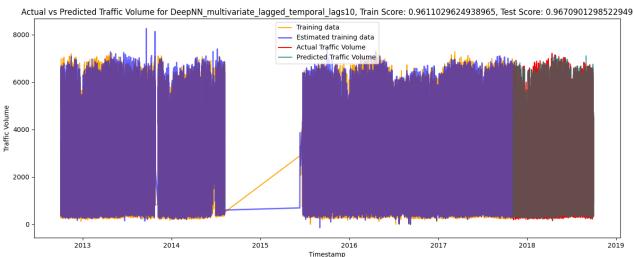
The best model for one\_step forecasting is the DeepNN model.

The model has been trained on the multivariate\_lagged\_temporal\_lags10 dataset.

The best score for the one\_step forecasting is 0.9670901298522949.

#### Best DeepNN forecast over time







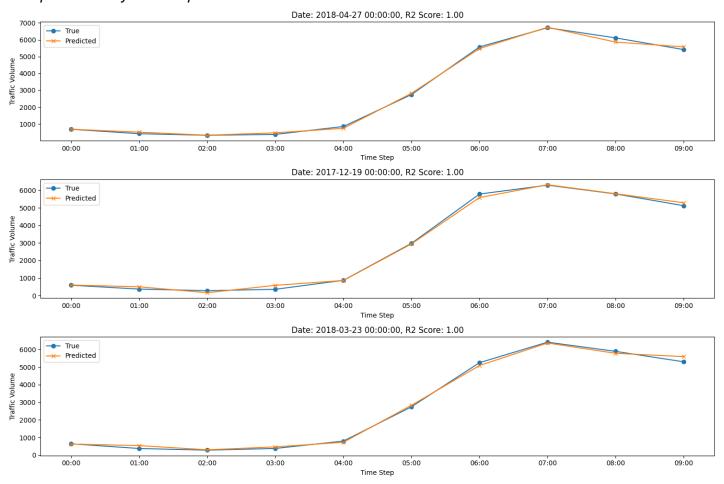
# The best model for multistep forecasting.

The best model for multistep forecasting is the DeepNN model.

The model has been trained on the multivariate\_lagged\_temporal\_multistep\_lags10\_steps10 dataset.

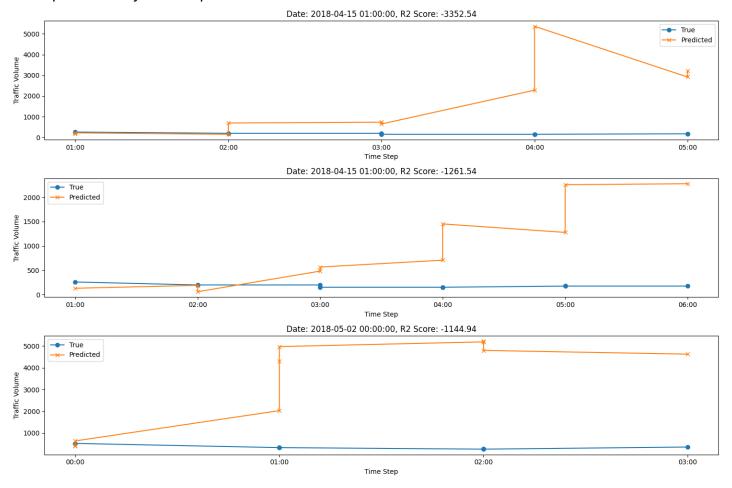
The best score for the multistep forecasting is 0.8332775235176086.

#### Best predicted days for DeepNN.





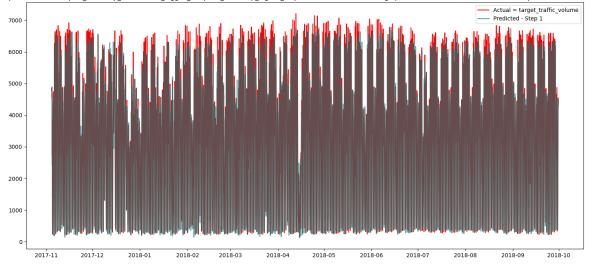
# Worst predicted days for DeepNN.



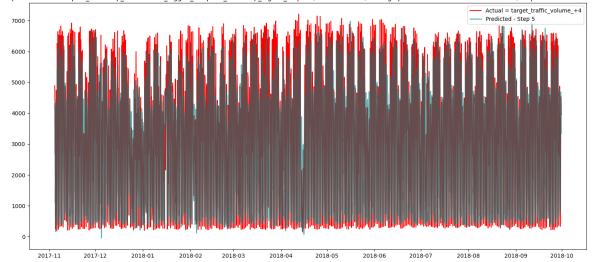


#### Steps plots for DeepNN forecasts over time

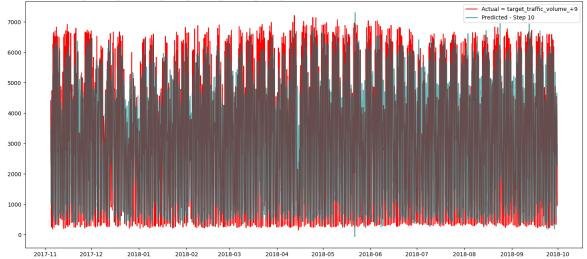








Multistep forecast for DeepNN\_multistep\_multivariate\_lagged\_temporal\_multistep\_lags10\_steps10, Traffic Volume, average performance 0.8332775235176086, step score 0.7077272630710357





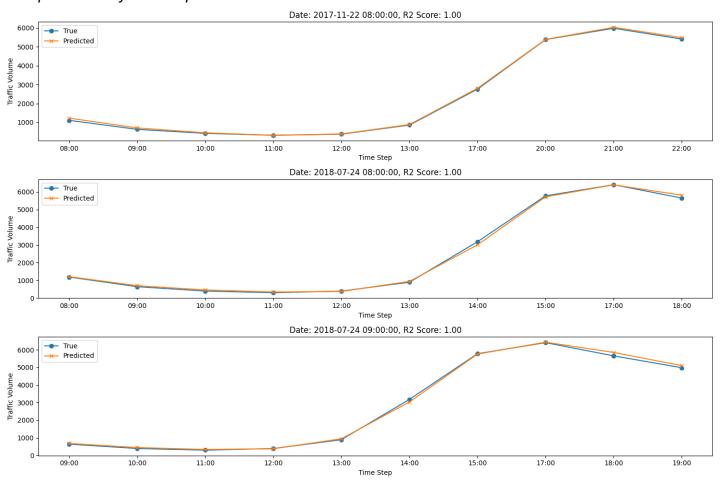
#### The best model for recursive forecasting.

The best model for recursive forecasting is the DeepNN model.

The model has been trained on the multivariate\_lagged\_temporal\_lags10 dataset.

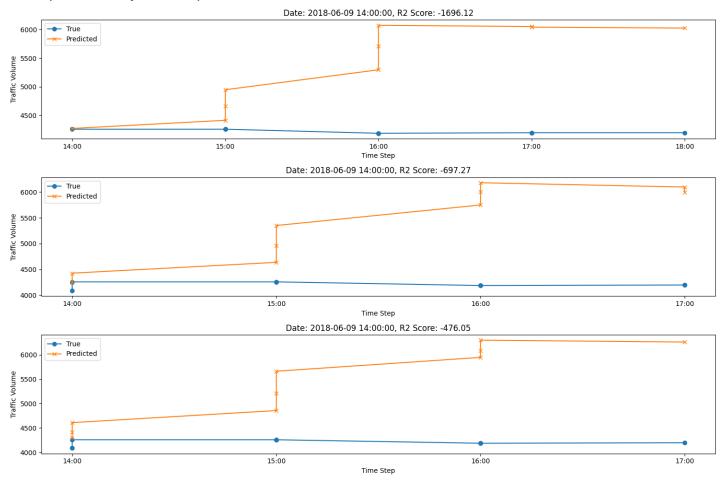
The best score for the recursive forecasting is 0.7752858400344849.

#### Best predicted days for DeepNN.





# Worst predicted days for DeepNN.





#### Steps plots for DeepNN forecasts over time





