From LSTM to Transformers: A Comprehensive Analysis of Deep Learning Techniques for Groundwater Contamination Forecasting at the Hanford Site

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Abstract

Keywords

1. Introduction

Hanford Site

Groundwater contamination

Climate change impact

Importance of AI-based solutions

Time-series forecasting importance

Different time-series models

Advantages of using DL-based models

Different types of DL models for time-series forecasting

Motivation of this manuscript

List of contributions of the manuscript

Reminder of the section

2. Background

Physics based models

Statistics-based models

DL-based models

3. Problem Formulation

Univariate forecasting

Multi-variate forecasting

Single step and multi-step forecasting

Sliding window – self supervised learning

Include definition of seq2seq

Timeseries characteristics (seasonality, trend) and Challenges

4. Deep learning models for time-series forecasting

4.1.1 RNN

4.1.2 LSTM

- 4.1.3 Bi-LSTM
- 4.1.4 GRU
- 4.1.5 Bi-GRU
- 4.1.6 CNN
- 4.1.7 TCN
- 4.1.8 Encoder-decoder
- 4.1.9 Attention
- 4.1.10 Transformer

DL architecture	Advantages	Disadvantages	Remarks

5. Results and analysis

5.1 Application area and dataset descriptions

Climate change data

Groundwater contamination data

5.2 Forecast error metrics

Mean Squared Error (MSE), Root Mean Squared Error (RMSE), Mean Absolute Error (MAE)

Coefficient of Determination

5.3 Performance analysis

Sequence length 4

Detect		IN	LSTM		Bi-LSTM		GRU		Bi-GRU		CNN		TCN		ED		Attention		MHA	
Dataset	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
Dataset1																				
Dataset2																				
Dataset3																				
Dataset4																				
Dataset5																				
Dataset6																				

Loss and accuracy convergence graphs.

Test error distributions

Model	Test	Dataset1		Dataset2		Data	set3	Data	set4	Data	set5	Dataset6	
Widdei	error	μ	σ	μ	σ	μ	σ	μ	σ	μ	σ	μ	σ
RNN													
LSTM													
BI_LSTM													
GRU													
BI_GRU													
CNN													
TCN													
ED													
Attention													
MHA													

Test error distribution graphs.	
Forecast comparison graphs.	
Confidence and prediction interval graphs.	

Pattern exploration using PCA.

Sequence length 50

Detect	RNN		LSTM		Bi-LSTM		GRU		Bi-GRU		CNN		TCN		ED		Attention		МНА	
Dataset	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE	MSE	MAE
Dataset1																				
Dataset2																				
Dataset3																				
Dataset4																				
Dataset5																				
Dataset6																				

Loss and accuracy convergence graphs.

Pattern exploration using PCA.

Test error distributions

Model	Test	Dataset		set1 Dataset2		Data	set3	Data	set4	Data	set5	Dataset6	
iviouei	error	μ	σ	μ	σ	μ	σ	μ	σ	μ	σ	μ	σ
RNN													
LSTM													
BI_LSTM													
GRU													
BI_GRU													
CNN													
TCN													
ED													
Attention													
Transformer													

Test error distribution graphs.	
Forecast comparison graphs.	
Confidence and prediction interval graphs.	

5.4 Analysis on model hyper-parameters
5.5 Frameworks and tools for timeseries forecasting
5.6 Recommendations for model section
6. Conclusion
References