

## Why use ELMs?

- Lightweight and easy to use algorithm for training Single Layer Feedforward Neural network (SFLNs).
- Randomised approach avoids bias problems.
- Lower training times than other traditional techniques like Back Propagation.
- It is a linear training process.

# Principle of Operation for ELM

- The basic training of ELM can be regarded as two steps: Random Initialization and Linear Parameter solution.
- ELM uses random parameters  $w_i$  and  $b_i$  in its hidden layer, and they are frozen during the whole training process.
- Moore-Penrose Inverse: A pseudo-inverse form of matrix which helps to compute the best fit solution to a system of linear equations which lacks a unique solution.

#### Training set: $S = [(x_i, y_i) | x_i \in \mathbb{R}^n, y_i \in \mathbb{R}^m, i = 1, ..., N]$ Initialization:

Assign random values to hidden weight  $w_i$  and bias  $b_i$  and calculate the output matrix of hidden layer H using training set.

#### **Analytical solution:**

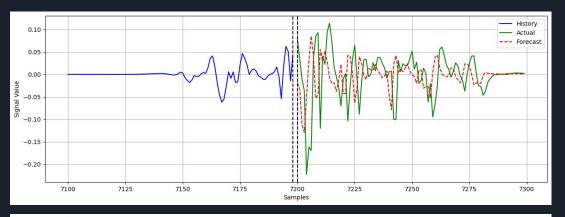
Obtain  $\beta$  from  $H\beta = T$  by Moore Penrose inverse.  $\beta = H^{\dagger}T$ , where  $H^{\dagger}$  is the Moore–Penrose generalized inverse of matrix H

### Results Obtained

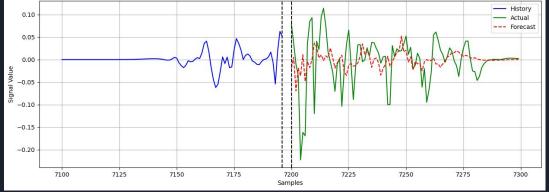
- The ELM Model was run for multiple time steps ahead ranging from 0 min (immediate) to 90 minutes ahead forecast horizons.
- The table below shows the results obtained:

Time (min)	RMSE	MAE	$\mathbf{R}^{2}$ (%)
0	0.0203	0.0063	78.0397
5	0.0284	0.0088	57.2428
15	0.0427	0.0157	3.8489
30	0.0441	0.0162	-2.4288
90	0.0411	0.0137	11.7849

## Forecast Plots



Immediate Forecast Horizon



90 minutes ahead Forecast Horizon

### Conclusion

- It can be observed that the ELM Model performs better than all the model used in this project.
- The randomised process helps to better generalise the complex time varying data.
- Single Layer of hidden neuron significantly cuts down the training time of the model.
- Future Work: ELMs can be combined and modified further to better suit different situations of machine learning problems.

Thank you