

Training Neural Networks

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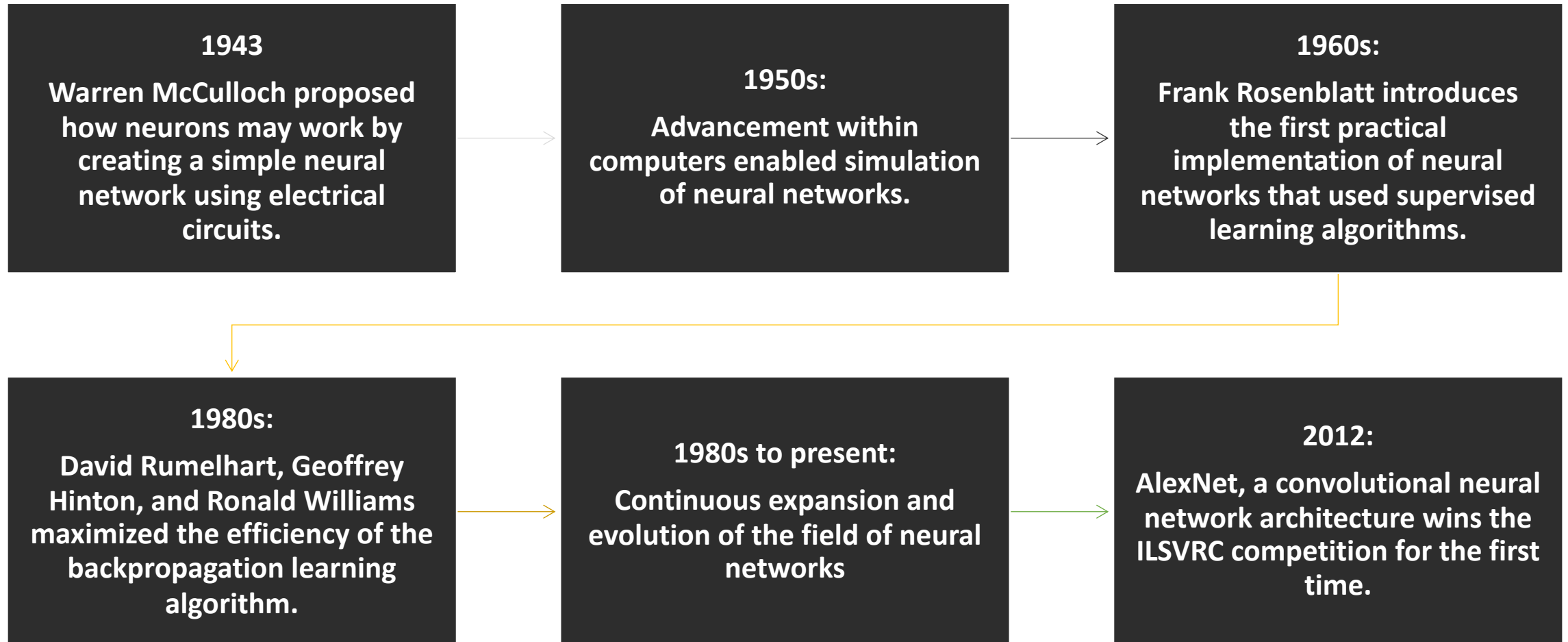
What is our project about?

Objective: Mathematical understanding of artificial neural networks, how to tune the parameters to improve the performance of each neural network.

Method: Train 3 types of Neural Networks:

- Feed-Forward Neural Networks
- Convolution Neural Networks
- Recurrent Neural Networks

Background



Applications

Forward-Feed Neural Networks (FNNs):

Fraud detection
Medical diagnosis
Handwriting recognition

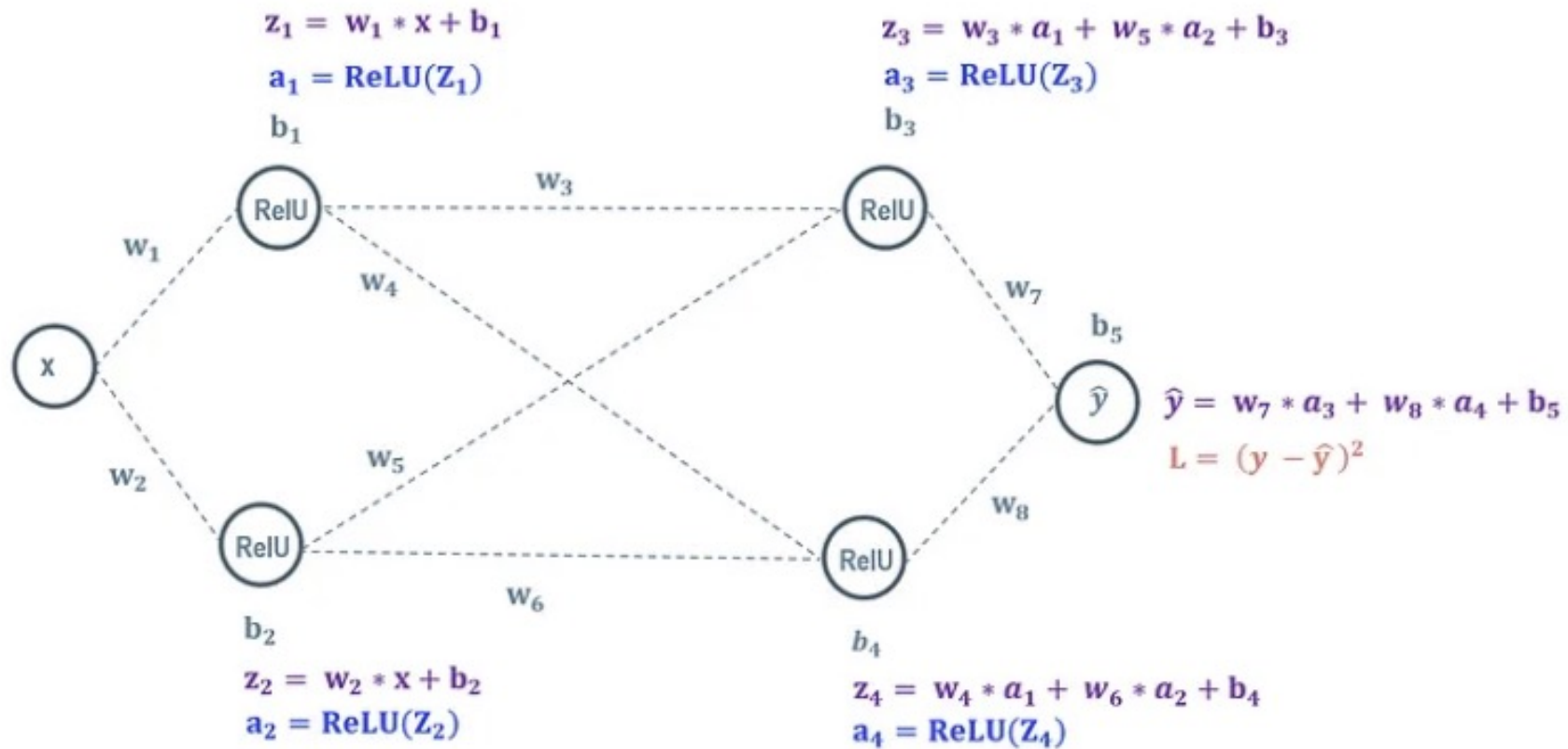
Convolutional Neural Networks (CNNs):

Image classification
Object detection and recognition
Facial recognition
Video analysis

Recurrent Neural Networks (RNNs):

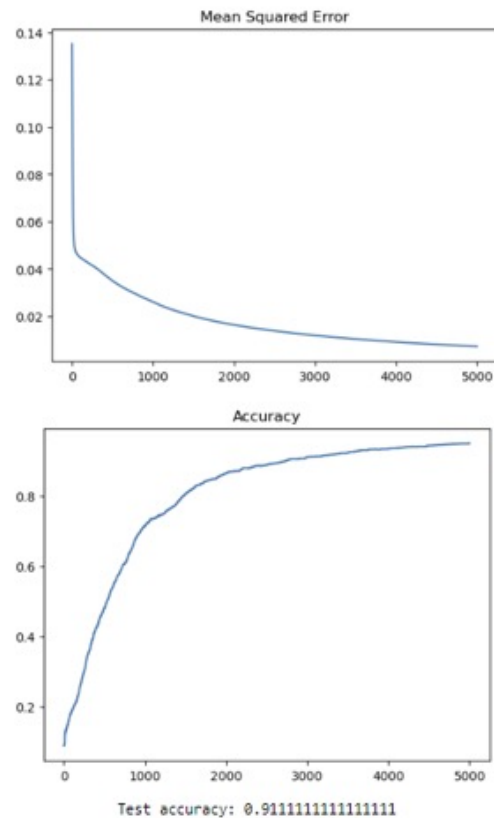
Prediction problems
Language modeling
Text generation
Speech recognition and translation
Forecasting

FNN Model & Architecture

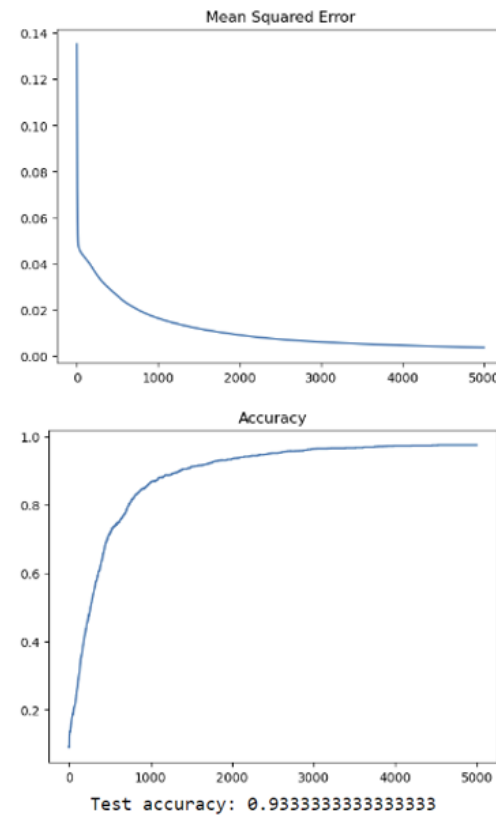


Findings for FNN Model

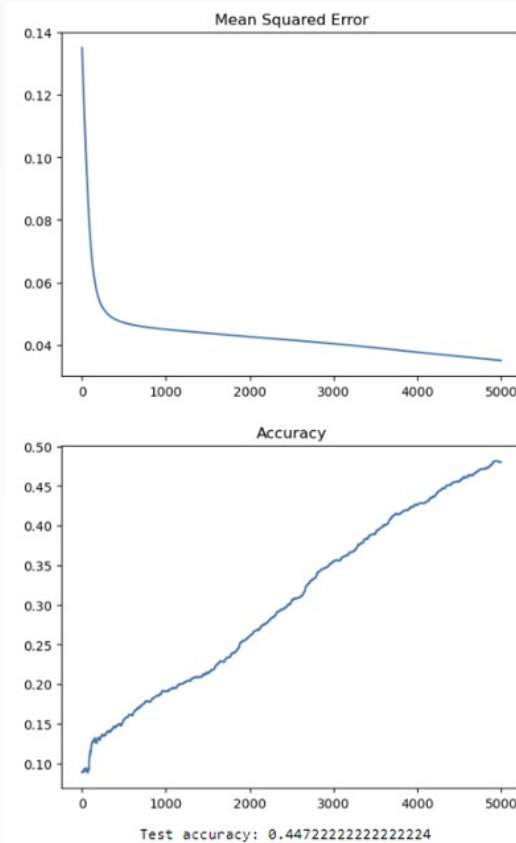
Learning Rate: 0.1
Iterations: 5000



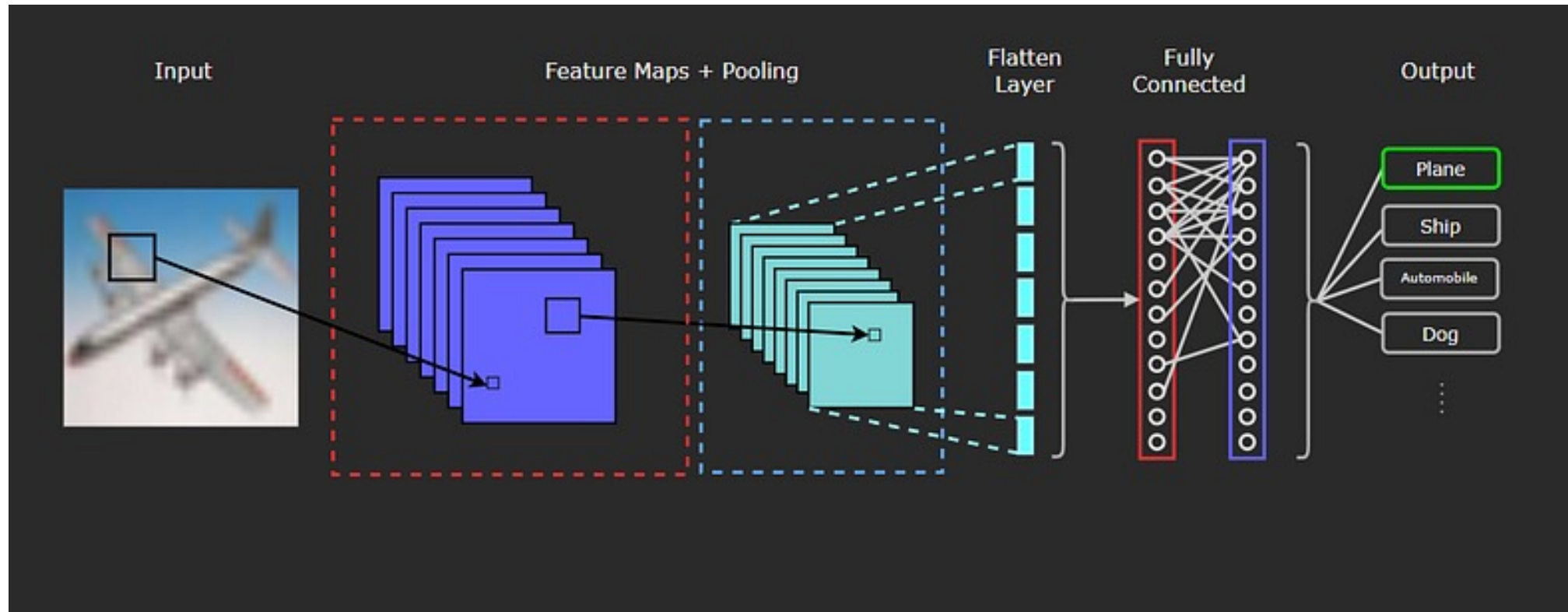
Learning Rate: 0.2
Iterations: 5000



Learning Rate: 0.01
Iterations: 5000

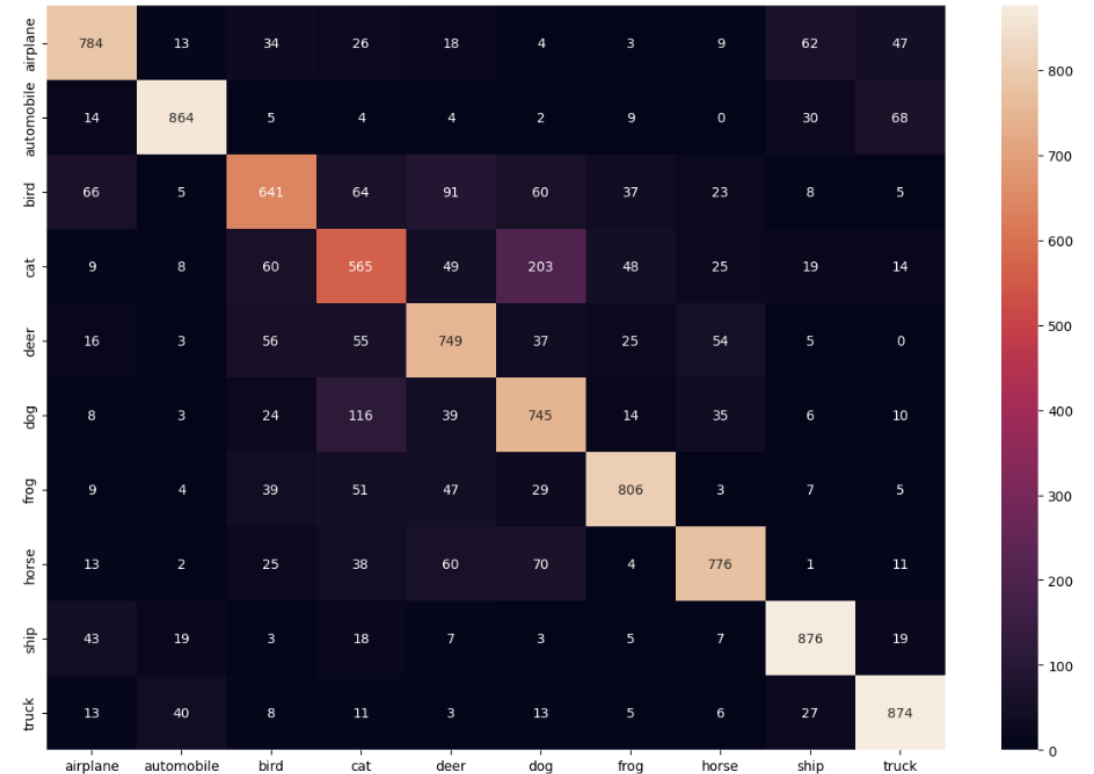


CNN Model & Architecture

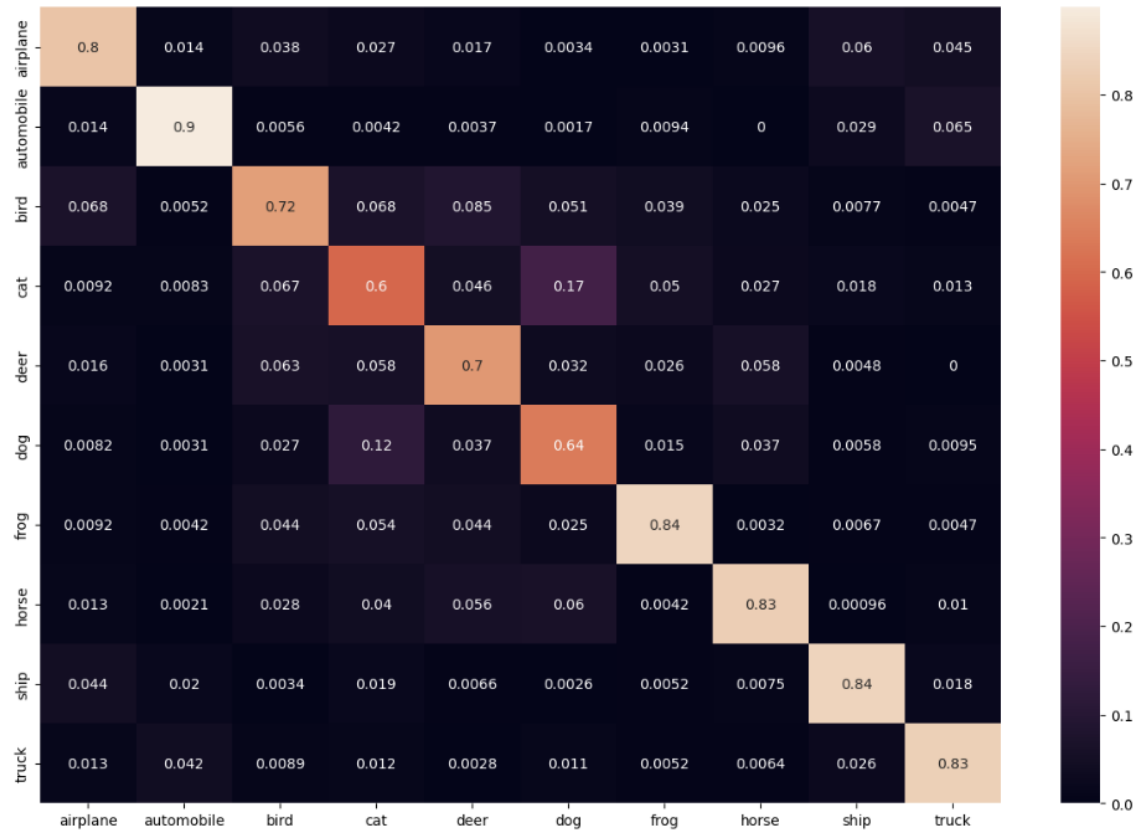


Findings for CNN Model

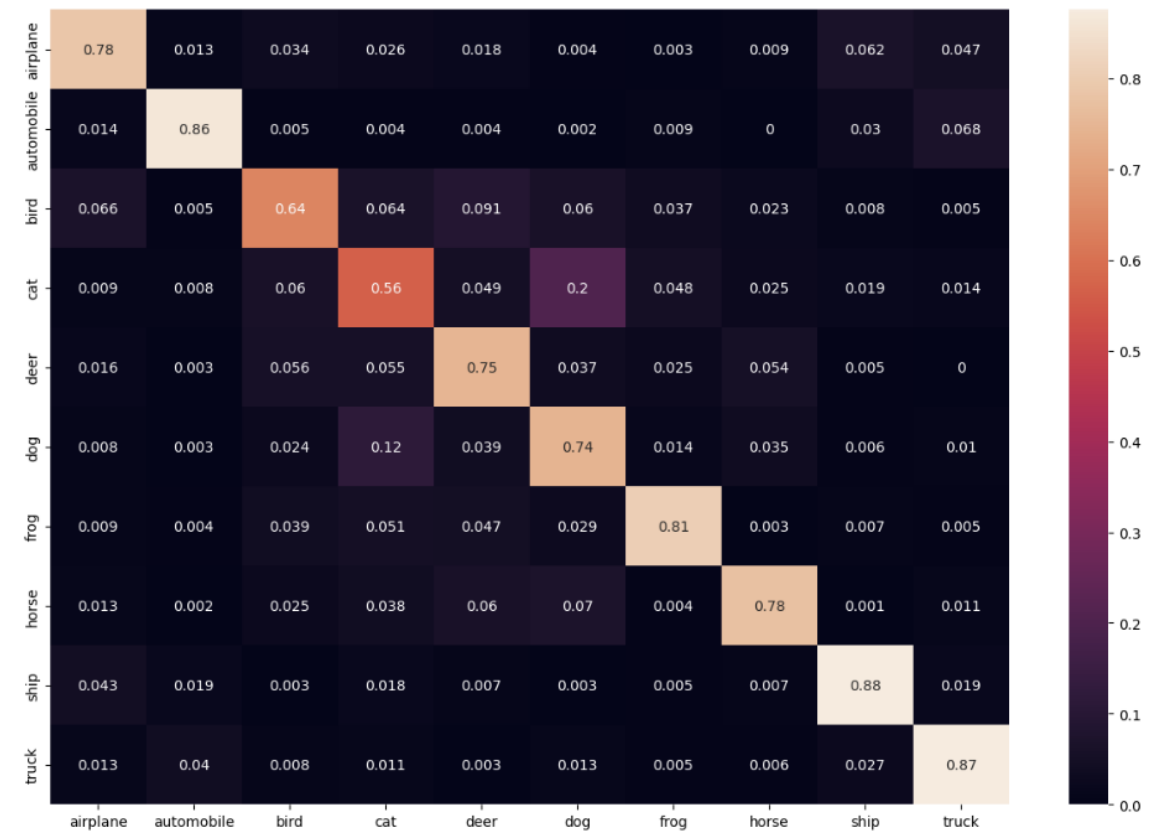
ACCURACY	
Airplane	0.882
Automobile	0.828
Bird	0.548
Cat	0.458
Deer	0.654
Dog	0.750
Frog	0.579
Horse	0.889
Ship	0.733
Truck	0.833
Avg Accuracy & Loss:	76.3% and 0.69 loss



Findings for CNN Model

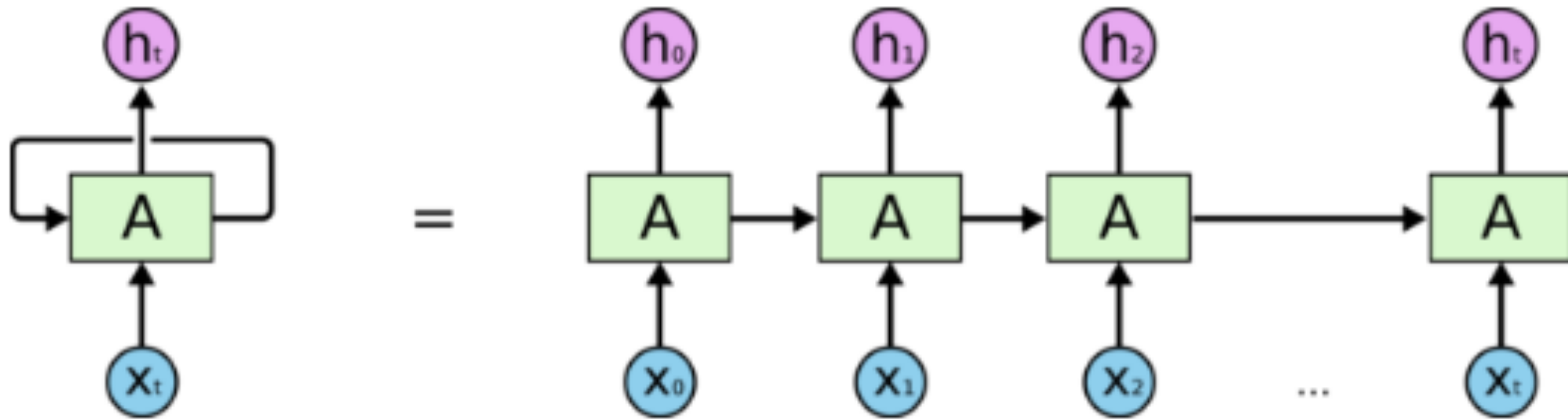


Precision: The fraction of identified positive cases correctly predicted



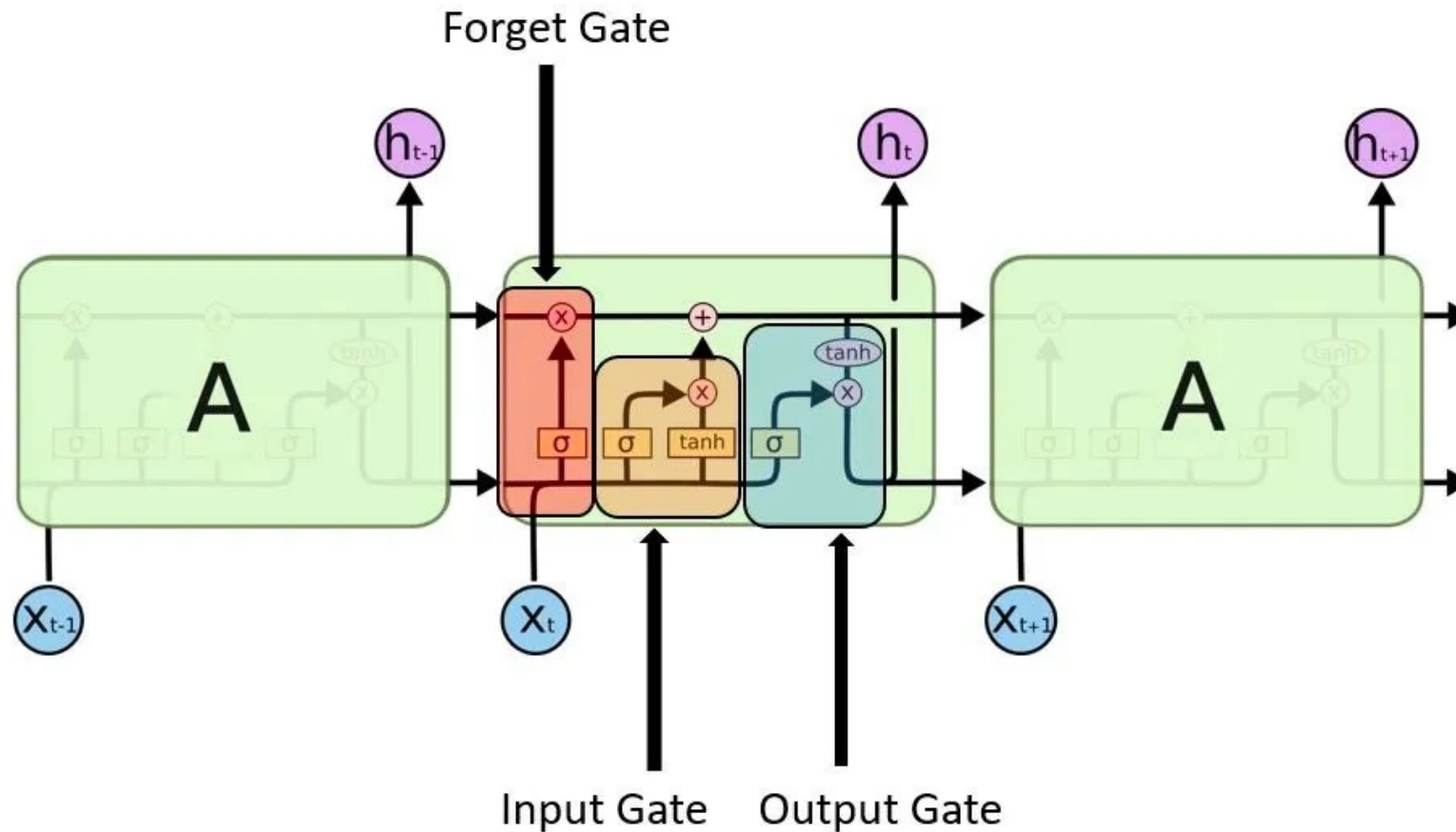
Recall: The fraction of actual positive cases that are correctly predicted

RNN Architecture



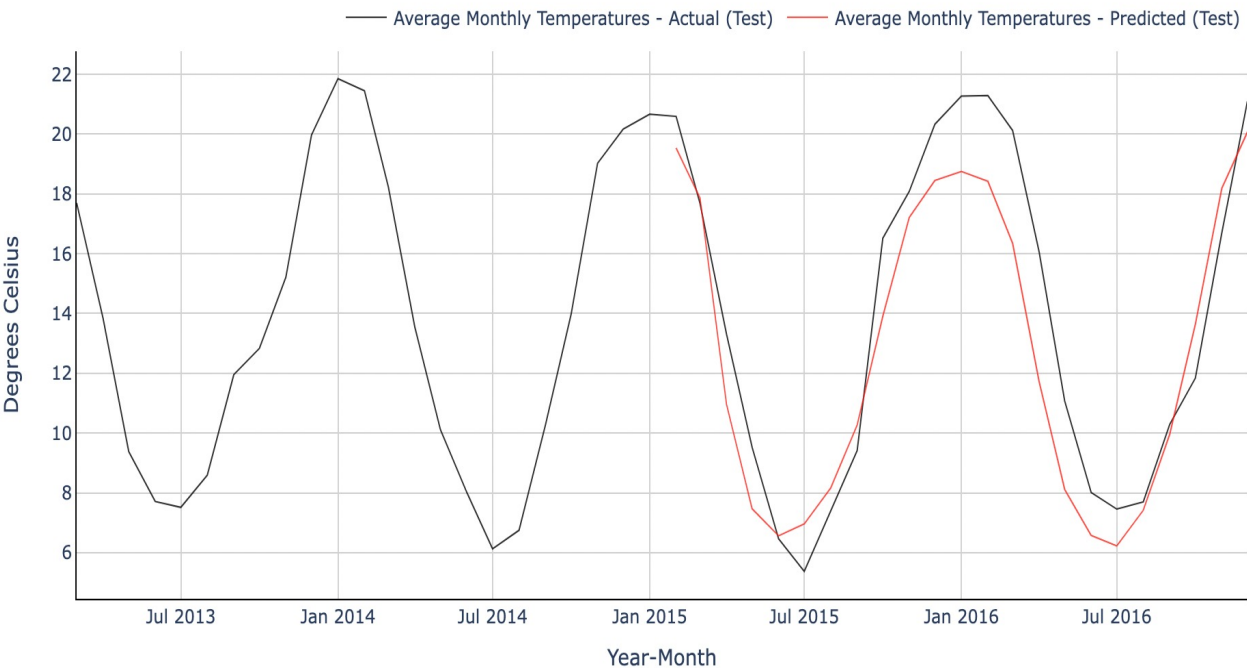
An unrolled recurrent neural network.

LSTM (Long Short Term Memory)



Findings for RNN Model

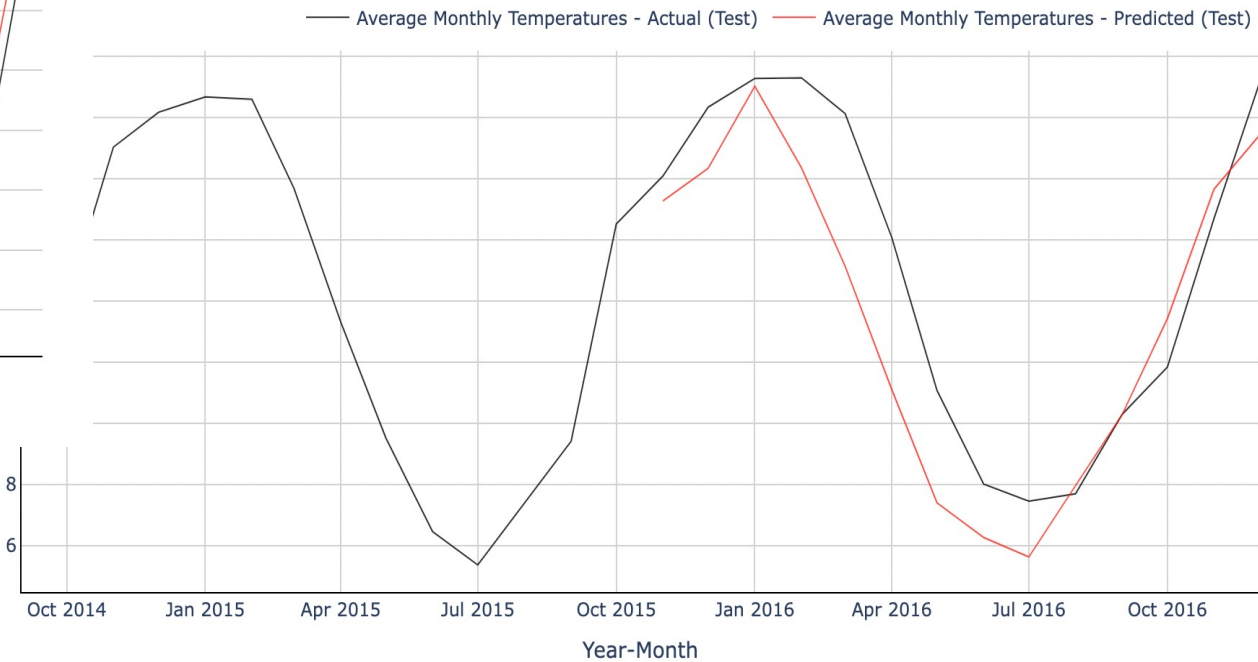
Average Monthly Temperatures



22 timestep + 1000 epoch

Canberra

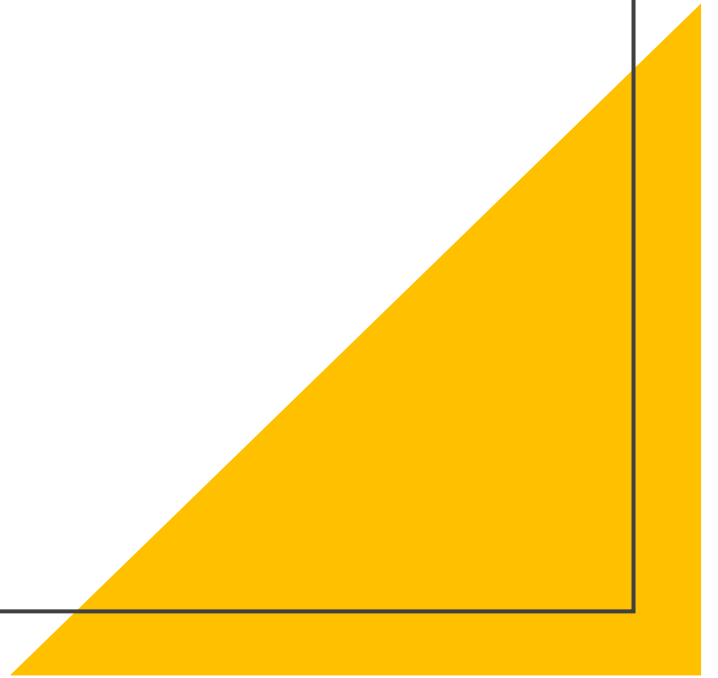
Monthly Temperatures



18 timestep + 2000 epoch

Future Work

Continue evaluating the mathematical concepts of our neural networks and determining how to best tune the hyperparameters of our model





Thank you!

References

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