# Lifetime Prediction of 1550 nm DFB Laser using **Machine Learning Techniques**



for  $h_i \geq$ 

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### Motivation

- The drawbacks of the conventional method:
  - Computationally complex and time consuming
  - Inaccurate and low level of confidence
  - Overestimation of actual MTTF

- Machine learning (ML) is a powerful tool to develop predictive maintenance approaches
- The ability of ML to glean insights from data within a complex and dynamic environment
- The ability of ML to adjust to new and unseen data

### **Objectives**

To develop a laser lifetime prediction model based on ML modelling the dependency between MTTF and the different laser parameters Accurate prediction of MTTF under different operating conditions

An artificial neural network (ANN) based approach predicts the **mean-time-to** failure (MTTF) of laser under different operating conditions with higher accuracy compared to the accelerated aging tests.

**ANN Model for Laser Lifetime Prediction** 

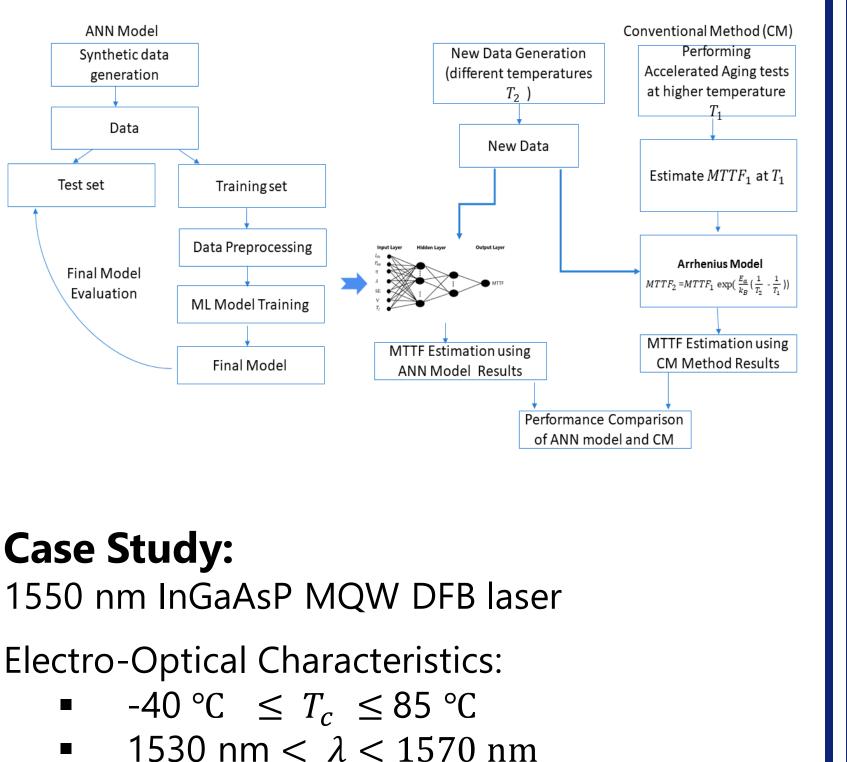
RMSE  $\mathsf{RMSE} = \sqrt{\frac{\sum (MTTF_{predicted} - MTTF_{GT})^2}{2}}$ Scoring Function S =  $\sum_{i=1}^{N} (e^{\frac{-n_i}{13}} - 1)$  for  $h_i < 0$  $\sum_{i=1}^{N} (e^{\frac{n_i}{10}} - 1).$ Where  $h_i \stackrel{0}{=} MTTF_{predicted} - MTTF_{GT}$ **ANN Model vs Conventional Method** 

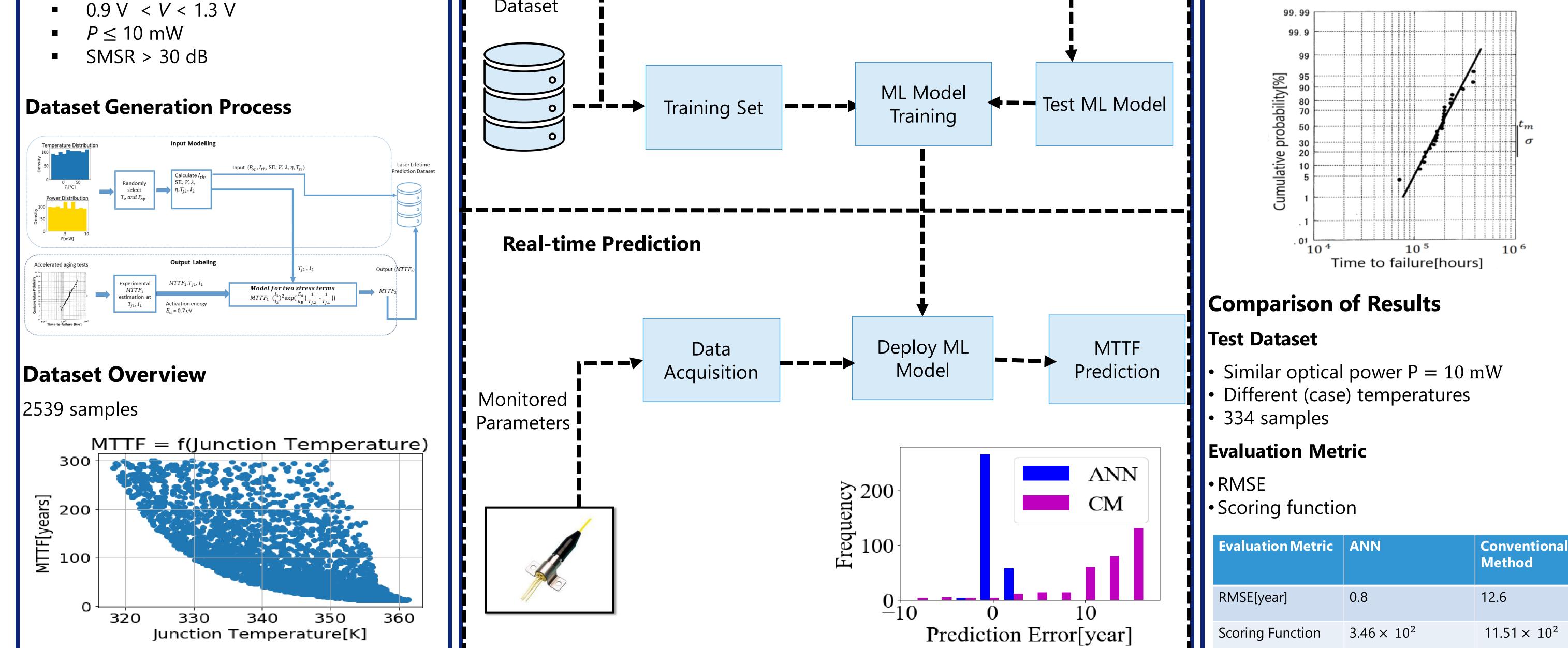
**Evaluation Metrics** 

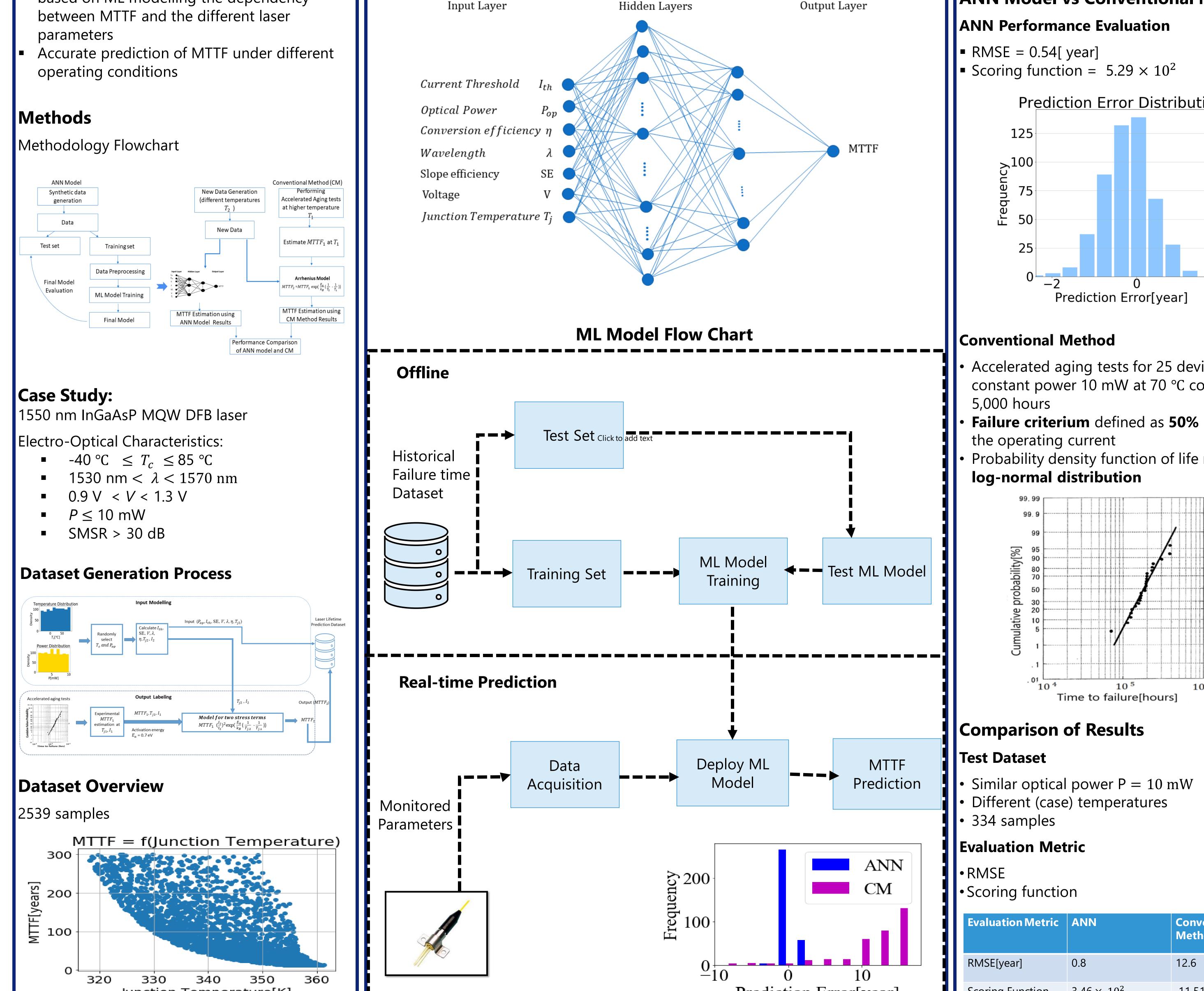
Results

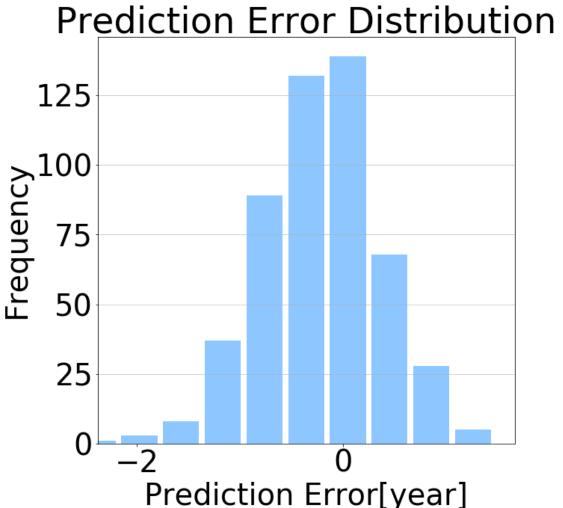
Input Layer

Hidden Layers









- Accelerated aging tests for 25 devices under constant power 10 mW at 70 °C conducted for
- Failure criterium defined as 50% increase of
- Probability density function of life modelled by

| 99, 99 r |   | <br>      |     |   |   |     | _ | _ |   | _ |
|----------|---|-----------|-----|---|---|-----|---|---|---|---|
| 99.99    | : | <br>      | 111 | 1 | : | : : |   | : | : | : |
|          | : | <br>: : : |     | : | : | : : |   | : | : | : |

### ANN Model

Two hidden layers:

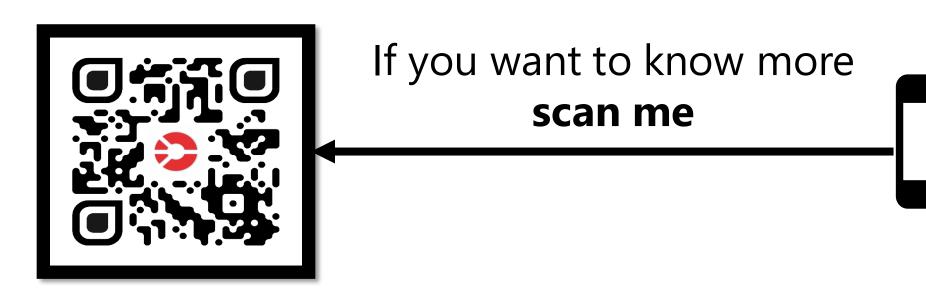
Ist hidden layer: 200 neurons 2nd hidden layer: 100 neurons

## Conclusions

- A novel approach for laser lifetime prediction using ANN has been presented
- The proposed ANN Model outperforms the conventional laser lifetime projection method
- Higher accuracy and applicability to unseen operating conditions

### **Next Steps**

 Collection of experimental or in-field data for the performance evaluation of the developed model



### References

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