

# Sandia Academic Alliance Fall 2021 University of Illinois LRD Virtual Poster Session

## Data-Driven Compact Modeling of Bipolar Junction Transistors with Recurrent Neural Networks



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### Introduction

**Objective:** Evaluate critically the capability of discrete-time RNN (DTRNN) to substitute for physics-based compact models in circuit simulation.

**Approach taken:** Start with a specific case study; compare the performance of a DTRNN model of an NPN Bipolar Junction Transistor (BJT) against that of the SPICE Gummel Poon (SGP) model.

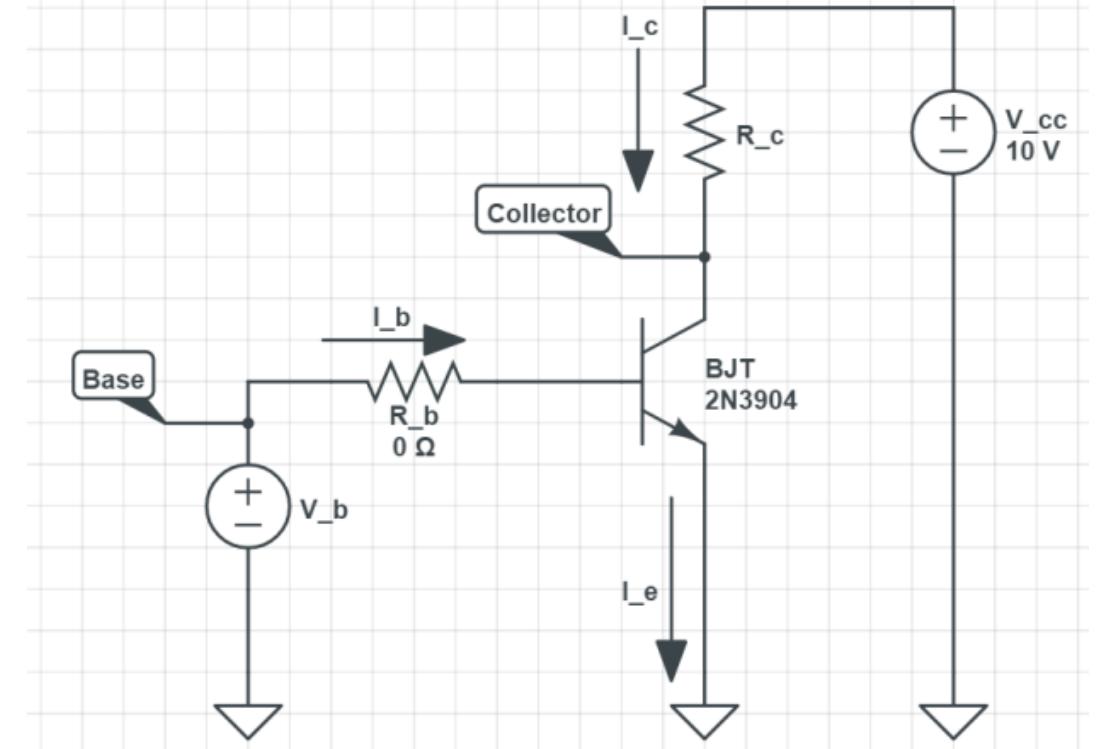
**Motivation:** Analysis and design of large-scale or complex circuits routinely involves compact semiconductor device models to reduce unnecessary computational burden in simulations and design optimization. Numerical, data-driven approaches demonstrate exciting potential to automate and accelerate compact model development and calibration.

### Approach

#### Experiment Details

This figure shows the setup used to generate training data. It includes two input voltages  $V_b$  and  $V_{cc}$ , a load resistor  $R_c$ , and the device-under-test (NPN model 2N3904).

$V_b$  and  $V_c$  denote the base and collector voltages, respectively.  $V_{cc}$  is a constant 10 V and  $R_b$  is fixed at 0 Ohms.  $V_b$  and  $R_c$  are varied.



#### RNN Model

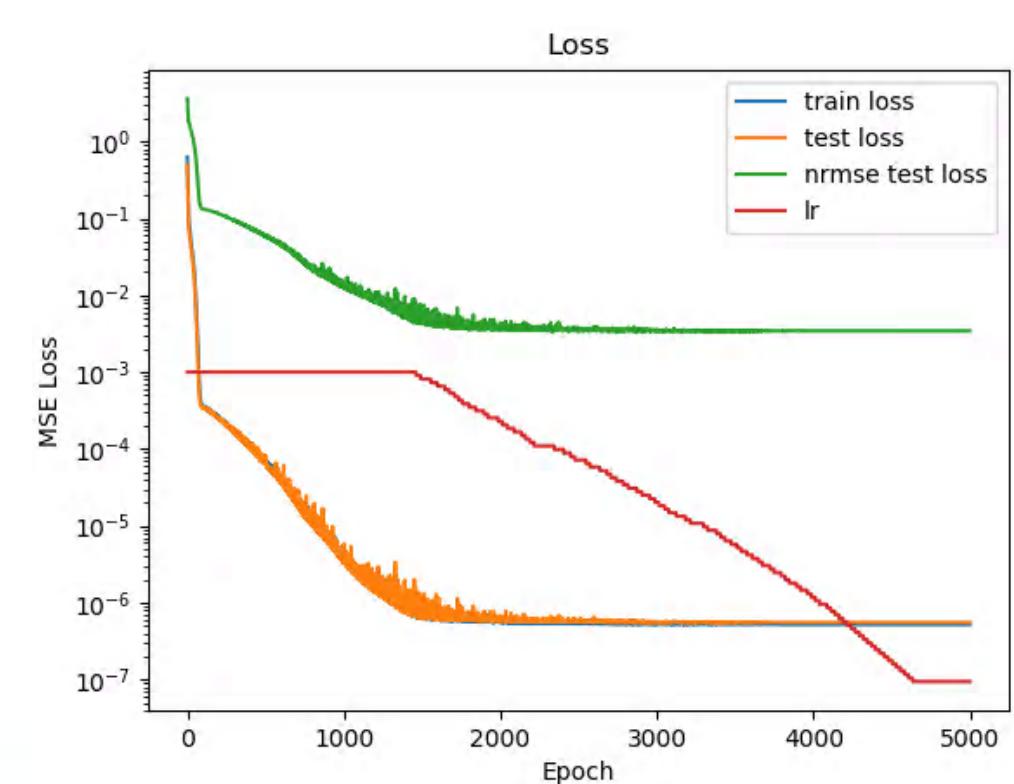
The inputs to the RNN will be  $V_b$  and  $V_c$  and the outputs will be  $I_b$  and  $I_c$ . We use a basic discrete-time RNN. A  $\tanh$  activation function is used, and the model output is taken from a linear layer.

$$h_t = \tanh(W_{ih}x_t + b_{ih} + W_{hh}h_{t-1} + b_{hh})$$

$$y = h_t A^T + b$$

**W** and **A** are weight matrices and **b** is a vector containing the bias terms. **x** are the model inputs. **h** are the hidden states. The dimension of **h** is a hyperparameter. The subscript indicates the time index.

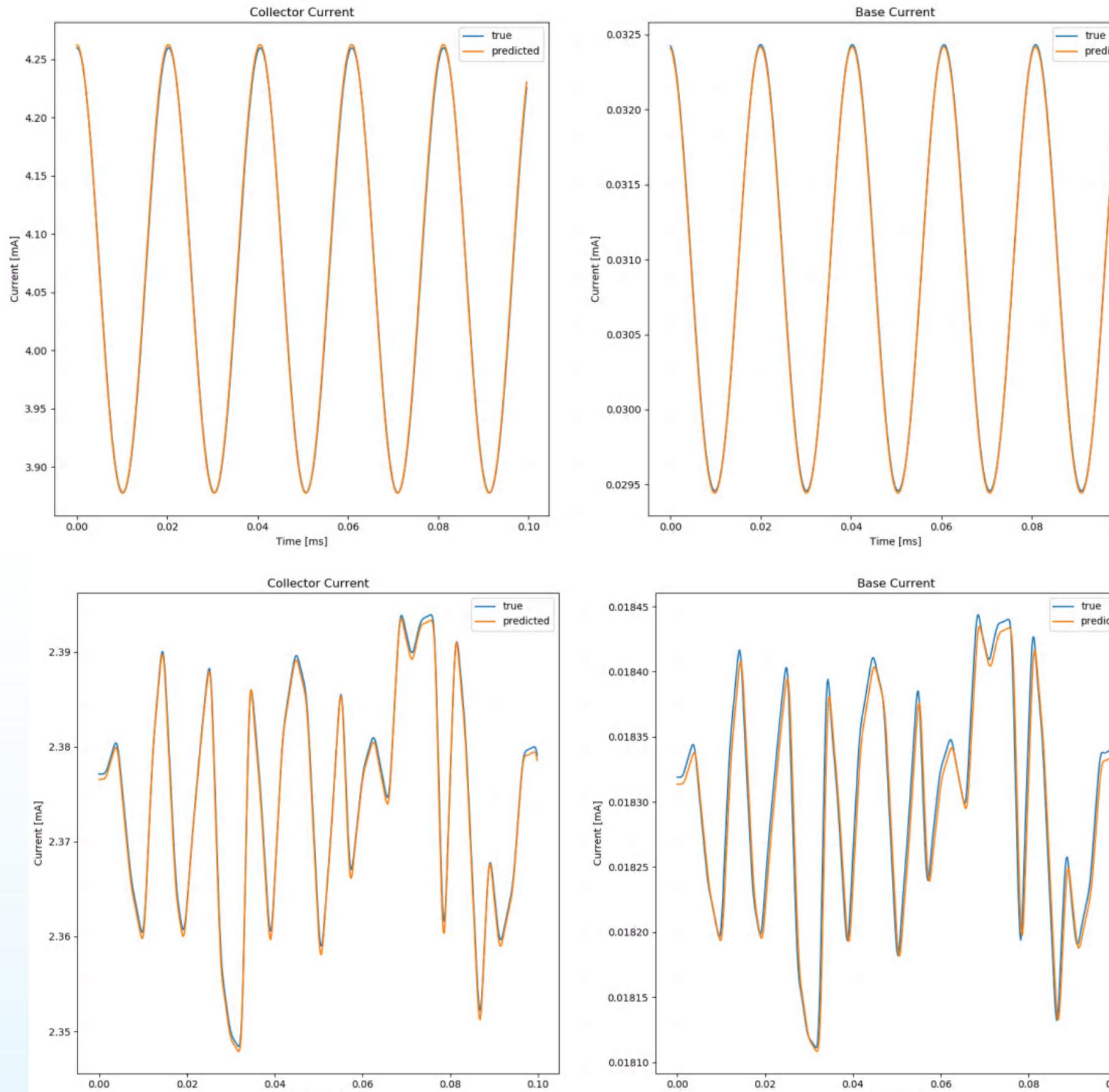
### Results



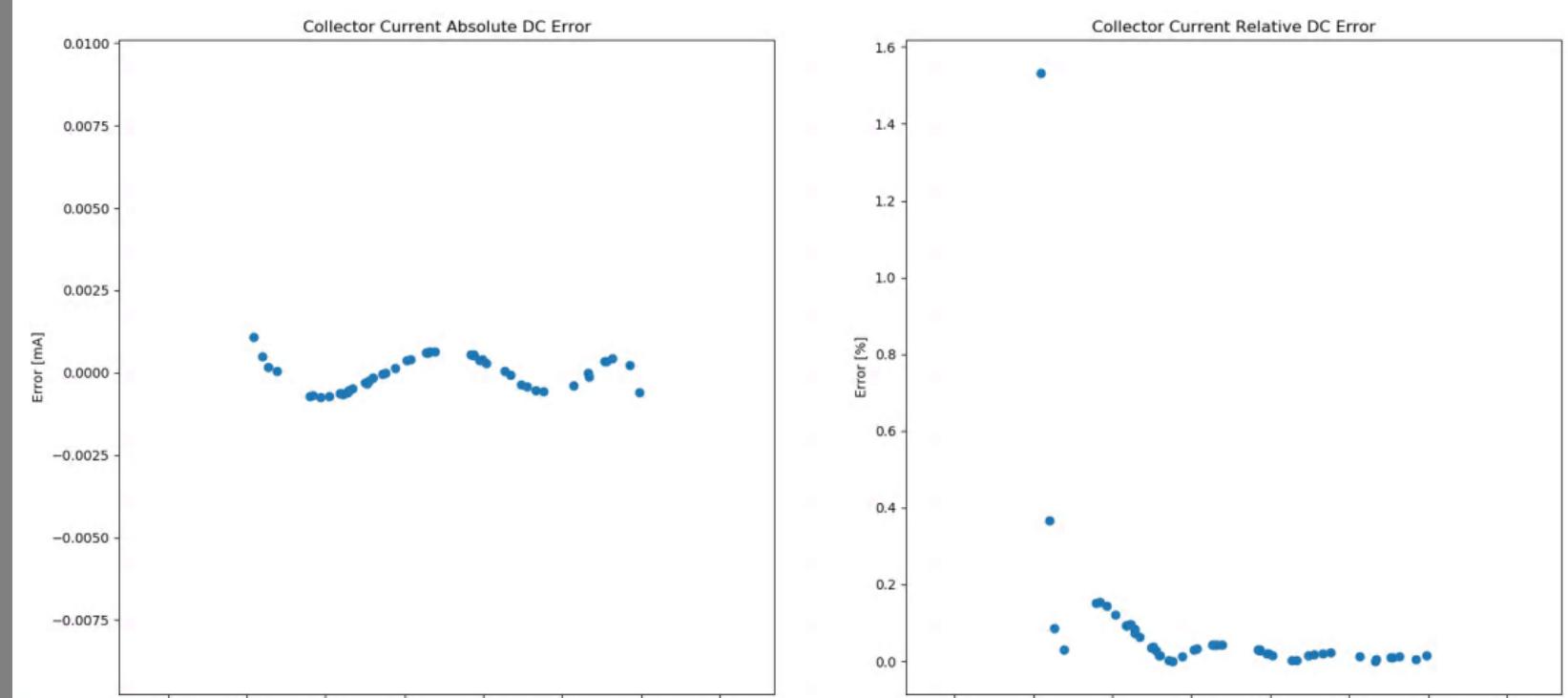
Top Left: Loss function for model trained and tested on sinusoidal samples.

Top Right: True (blue) and predicted (orange) currents for a sinusoidal waveform.

Bottom Right: True (blue) and predicted (orange) currents for a piecewise linear waveform.



### Analysis



- The DTRNN model achieves a small loss when the load resistor is constant.
- The figure shows absolute and relative DC errors vs.  $V_b$  for a model trained and tested on sinusoidal data.
- The average relative error is 0.1%. However, the absolute DC error, while very small, has a distinct pattern that requires further investigation.

### Conclusion

It is challenging to optimize a DTRNN to represent both the large signal transient response of the device and its small signal AC response with comparable accuracy.

The results suggest that a DTRNN may be a useful temporary model that is used to represent an emerging device with reasonable accuracy until the physics-based model can be developed.