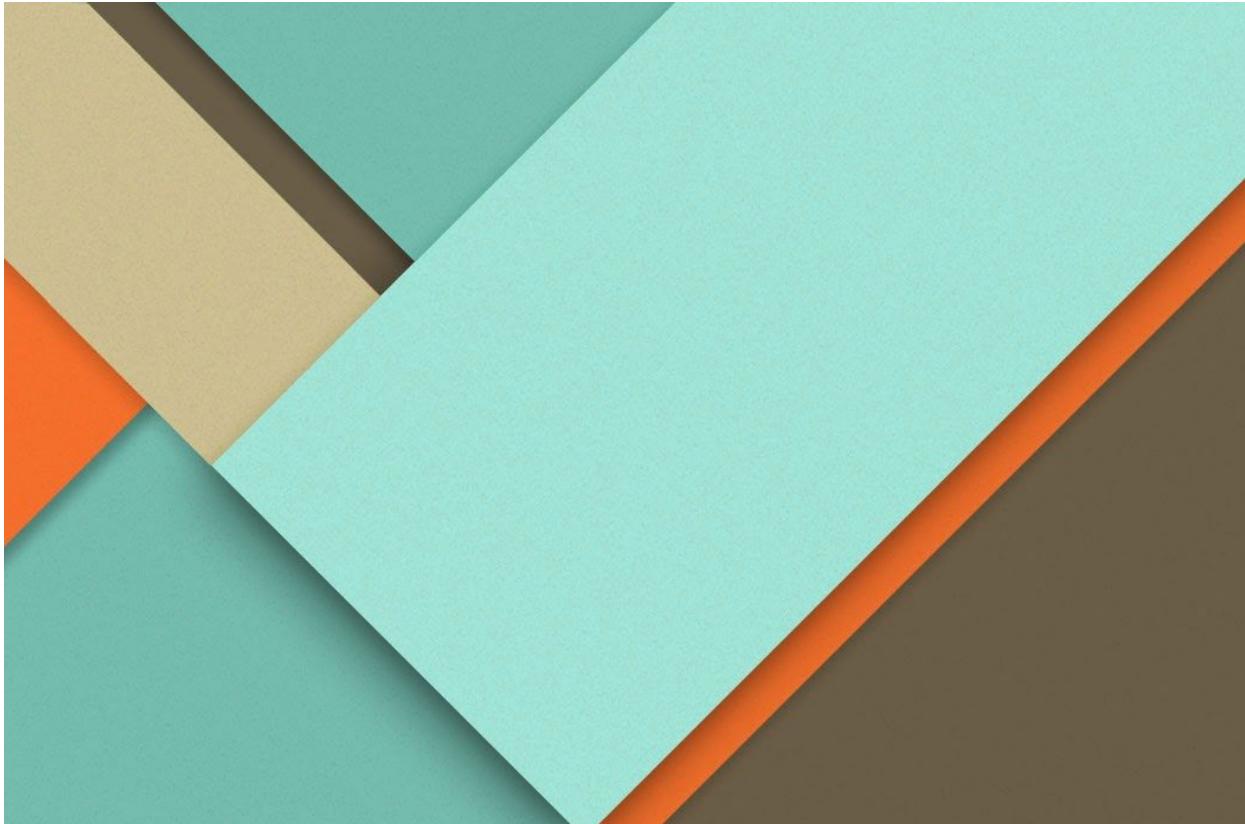


# Team Mechismu Racing

# Tyre Data Analysis Using Deep Learning

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Slip Rate Prediction using Keras API using a given dataset

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—TDA\_Build1 Models

## Overview

Here is a report on the very first attempt of my approach towards predicting the SR (Slip Rate) values from our indigenous tyre dataset with 20 features. Tensorflow 2.0 's Keras API was used to create a regression model for the same.

### **USED:**

*Google Colab*

*Tensorflow Version 2.3.0*

*Keras Version 2.4.0*

*Python 3.6+*

## About the dataset

**Dataset used :** B1654raw48.dat

- The dat file was accessed on MS Excel and then saved as .csv(comma delimited).

### **Dataset Description:**

21 columns (Features)

44901 rows (Instances)

### **Features Inputs (20 items):**

ET V N SA IA RL RE P FX FY FZ MX MZ NFX NFY RST TSTI TSTC TSTO AMBTMP

### **Label Outputs (1 item to predict):**

SR



### **Dataset Preparation(Making it trainable):**

- The dataset was normalized manually but was not a hopeful approach
- In the first Build version model MinMaxScaler() was used which has shown some promising results
- 20% of the full dataset was splitted into Trainfull(80% of total) and Test sets(20%)
- 20% of the Trainfull Set was further splitted into Train(80% of Trainfull) and validation sets(20%)

### **Model Structure(Build\_1 models):**

#### **Layers(Total 6 Layers)**

- Input Layer
- 4 Hidden Layers
- Output Layer

#### **Neurons Per layer:**

1000->1000->500->500->250->1

**The models are “normal” initialized and L2 regularized**

**(It was discovered that this formation worked better)**

#### **Activations:**

Relu, Linear(output)

#### **Optimizers:**

Nadam, Adam

#### **Patience:**

10-30

#### **Loss function + metrics:**

MAE



## ***What produced a variety of Build\_1 models ?***

Tweaking of:

- L2 rate
- Learning Rate
- Patience
- Selection of Optimizers
- Epochs

Created a variety of Build\_1 models

## **Milestones**

### **I. TDA\_build1\_1**

Fully Accurate Predictions:

6592

Predictions with an error of +/- 0.001:

2389

Predictions with an error more than +/- 0.001:

0

### **II. TDA\_build1\_2**

Fully Accurate Predictions:

7969

Predictions with an error of +/- 0.001:

1012

Predictions with an error more than +/- 0.001:

0

### **III. TDA\_build1\_3**

Fully Accurate Predictions:



5500

Predictions with an error of +/- 0.001:

3481

Predictions with an error more than +/- 0.001:

0

#### IV. TDA\_build1\_4

Fully Accurate Predictions:

6590

Predictions with an error of +/- 0.001:

2387

Predictions with an error more than +/- 0.001:

4

#### V. TDA\_build1\_5

Fully Accurate Predictions:

7903

Predictions with an error of +/- 0.001:

1078

Predictions with an error more than +/- 0.001:

0

## Goals

1. Planning to increase the number of layers and neurons in the TDA\_build2 models to see if they perform better. Along with BatchNormalization.
2. Also Planning to use SELU, ELU, Leaky Relu, Gelu in the TDA\_build3 models to see if they perform better.
3. Might shift to decision trees and other conventional regression algorithms in case the model performs better in those.

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4. I'll produce a final python program with proper EDA of the dataset after the final model and final purpose of prediction is fixed. This is just a basic model described to showing and suggesting a use of Machine Learning in tyre data analysis