Import the libraries

```
In [1]: #importing the libraries
  import tensorflow as tf
  from tensorflow import keras
  import numpy as np
  import pandas as pd
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losses .sparse_softmax_cross_entropy instead.

Load the data and data pre-processing

```
baseline df = pd.read excel('extracted features baseline.xlsx')
        toolwear df = pd.read excel('extracted features toolwear.xlsx')
In [3]: #labelling the datasets. O for baseline, 1 for toolwear. This will be the variable the model tries to predict
        baseline_df["state"] = 0
        toolwear_df["state"] = 1
In [4]: #concantanate the datasets
        combined df = pd.concat([baseline df, toolwear_df], axis=0)
        print(combined_df.shape)
       (840, 67)
In [5]: #getting the y label
        state = combined_df["state"].values
        print(state.shape)
       (840,)
In [6]: #getting the features to train the model
        features = combined_df.drop('state', axis=1).values
        print(features.shape)
       (840, 66)
In [7]: #train test split
        from sklearn.model selection import train test split
        X train, X test, Y train, Y test = train test split(features, state, test size=0.2, random state=50)
In [8]: #data scalling
        from sklearn.preprocessing import StandardScaler
        sc = StandardScaler()
        X_train = sc.fit_transform(X train)
        X_test = sc.transform(X_test)
```

Constructing the stacked autoencoder and training it

```
condensed e1 = 50
         condensed e2 = 30
         condensed_e3 = 10
In [10]: #constructing the model
         #input layer which number of neurons equals the number of original features
         l_in = keras.Input(features.shape[1])
         #hidden layer of encoder 1
         l_e1 = keras.layers.Dense(condensed_e1)(l_in)
         #hidden layer of encoder 2
         l_e2 = keras.layers.Dense(condensed_e2)(l_e1)
         #hidden layer of encoder 3
         l_e3 = keras.layers.Dense(condensed_e3)(l_e2)
         #hidden layer of decoder 2
         l d2 = keras.layers.Dense(condensed e2)(l e3)
         #hidden layer of decoder 1
         l d1 = keras.layers.Dense(condensed e1)(l d2)
```

In [9]: #specify the number of condensed features for the 3 encoders. This will be the number of neurons in the hidden

```
#output layer which is the same as the input
l_out = keras.layers.Dense(features.shape[1])(l_d1)
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\backend.py:1398: The name tf.executing_eagerly_outside_functions is deprecated. Please use tf.compat.v1.executing eagerly outside functions instead.

```
In [11]: #defining the autoencode
autoencoder = keras.Model(l_in, l_out)
```

In [12]: autoencoder.summary()

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 66)]	0
dense (Dense)	(None, 50)	3350
dense_1 (Dense)	(None, 30)	1530
dense_2 (Dense)	(None, 10)	310
dense_3 (Dense)	(None, 30)	330
dense_4 (Dense)	(None, 50)	1550
dense_5 (Dense)	(None, 66)	3366

Total params: 10436 (40.77 KB) Trainable params: 10436 (40.77 KB) Non-trainable params: 0 (0.00 Byte)

```
In [13]: #compile the model
autoencoder.compile(optimizer='adam', loss='mse')
#train the model
autoencoder.fit(X_train, X_train, epochs = 50, batch_size = 8, validation_split = 0.1)
```

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\optimizers__init__.py:309: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimize r instead.

Epoch 1/50

WARNING:tensorflow:From C:\Users\Teo Boon Kean\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\s rc\utils\tf_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecated. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

```
76/76 [============= ] - 6s 14ms/step - loss: 0.7300 - val loss: 0.3674
Epoch 2/50
76/76 [=============] - 0s 5ms/step - loss: 0.3338 - val loss: 0.2611
Epoch 3/50
Epoch 4/50
76/76 [=====
       Epoch 5/50
76/76 [====
             ========] - 0s 5ms/step - loss: 0.1864 - val_loss: 0.1638
Epoch 6/50
76/76 [====
             ========] - Os 6ms/step - loss: 0.1709 - val_loss: 0.1560
Epoch 7/50
76/76 [=====
           ========] - 0s 6ms/step - loss: 0.1620 - val_loss: 0.1505
Epoch 8/50
Epoch 9/50
76/76 [=============] - 0s 5ms/step - loss: 0.1530 - val loss: 0.1422
Epoch 10/50
76/76 [============= ] - 0s 5ms/step - loss: 0.1512 - val loss: 0.1431
Epoch 11/50
Epoch 12/50
          76/76 [=====
Epoch 13/50
76/76 [==========] - 0s 6ms/step - loss: 0.1480 - val loss: 0.1392
Epoch 14/50
76/76 [============] - 0s 5ms/step - loss: 0.1475 - val_loss: 0.1395
Epoch 15/50
76/76 [============= ] - 0s 6ms/step - loss: 0.1470 - val loss: 0.1406
Epoch 16/50
Epoch 17/50
```

```
Epoch 18/50
    Epoch 19/50
    76/76 [==
                =======] - Os 5ms/step - loss: 0.1465 - val loss: 0.1393
    Epoch 20/50
    76/76 [=====
                =======] - 0s 6ms/step - loss: 0.1461 - val loss: 0.1363
    Epoch 21/50
    76/76 [=====
            Epoch 22/50
    76/76 [=====
            Epoch 23/50
    Epoch 24/50
    76/76 [========== ] - 0s 5ms/step - loss: 0.1463 - val loss: 0.1366
    Epoch 25/50
    76/76 [========== ] - 0s 5ms/step - loss: 0.1460 - val loss: 0.1389
    Epoch 26/50
    Epoch 27/50
    76/76 [======
            Epoch 28/50
    76/76 [=====
               ========] - 0s 5ms/step - loss: 0.1461 - val loss: 0.1386
    Epoch 29/50
    76/76 [====
              =========] - Os 5ms/step - loss: 0.1469 - val loss: 0.1384
    Epoch 30/50
    76/76 [======
            Epoch 31/50
    Epoch 32/50
    76/76 [============= ] - 0s 5ms/step - loss: 0.1454 - val loss: 0.1371
    Epoch 33/50
    Epoch 34/50
    76/76 [=====
              =========] - Os 5ms/step - loss: 0.1455 - val loss: 0.1372
    Epoch 35/50
    76/76 [============] - 0s 5ms/step - loss: 0.1452 - val loss: 0.1371
    Epoch 36/50
           76/76 [=====
    Epoch 37/50
    76/76 [=====
              =========] - 0s 5ms/step - loss: 0.1453 - val_loss: 0.1366
    Epoch 38/50
    76/76 [======
           Fnoch 39/50
    Epoch 40/50
    Epoch 41/50
    Epoch 42/50
    76/76 [=====
              ========] - Os 6ms/step - loss: 0.1463 - val_loss: 0.1376
    Epoch 43/50
    76/76 [=====
               ========] - Os 6ms/step - loss: 0.1461 - val loss: 0.1380
    Epoch 44/50
    76/76 [=====
           Epoch 45/50
    Epoch 46/50
    Epoch 47/50
    76/76 [============ ] - 0s 5ms/step - loss: 0.1459 - val loss: 0.1363
    Epoch 48/50
    76/76 [============= ] - 0s 6ms/step - loss: 0.1458 - val loss: 0.1374
    Fnoch 49/50
    Epoch 50/50
    76/76 [============] - 0s 5ms/step - loss: 0.1452 - val_loss: 0.1389
Out[13]: <keras.src.callbacks.History at 0x20630cb3ed0>
```

Model Evaluation

```
In [14]: from sklearn.metrics import mean_absolute_error

#Mean square error of the model
pred = autoencoder.predict(X_test)
print(mean_absolute_error(X_test,pred))
```

6/6 [=======] - 0s 4ms/step 0.27588827367051055