```
#importing libraries and dataset
import numpy as np
import pandas as pd
import tensorflow as tf
import matplotlib.pyplot as plt
from sklearn.metrics import accuracy_score
from tensorflow.keras.optimizers import Adam
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras import Model, Sequential
from tensorflow.keras.layers import Dense, Dropout
from sklearn.model_selection import train_test_split
from tensorflow.keras.losses import MeanSquaredLogarithmicError

PATH_TO_DATA = 'http://storage.googleapis.com/download.tensorflow.org/data/ecg.csv'
data = pd.read_csv(PATH_TO_DATA, header=None)
data.head()
```

	0	1	2	3	4	5	6	7	8	9	...	131	132	133	134	135	
0	-0.112522	-2.827204	-3.773897	-4.349751	-4.376041	-3.474986	-2.181408	-1.818286	-1.250522	-0.477492	...	0.792168	0.933541	0.796958	0.578621	0.257740	0.2
1	-1.100878	-3.996840	-4.285843	-4.506579	-4.022377	-3.234368	-1.566126	-0.992258	-0.754680	0.042321	...	0.538356	0.656881	0.787490	0.724046	0.555784	0.4
2	-0.567088	-2.593450	-3.874230	-4.584095	-4.187449	-3.151462	-1.742940	-1.490659	-1.183580	-0.394229	...	0.886073	0.531452	0.311377	-0.021919	-0.713683	-0.5
3	0.490473	-1.914407	-3.616364	-4.318823	-4.268016	-3.881110	-2.993280	-1.671131	-1.333884	-0.965629	...	0.350816	0.499111	0.600345	0.842069	0.952074	0.9
4	0.800232	-0.874252	-2.384761	-3.973292	-4.338224	-3.802422	-2.534510	-1.783423	-1.594450	-0.753199	...	1.148884	0.958434	1.059025	1.371682	1.277392	0.9

5 rows × 141 columns

```
#finding shape of the dataset
data.shape
```

```
(4998, 141)
```

```
#splitting training and testing dataset
features = data.drop(140, axis=1)
target = data[140]
x_train, x_test, y_train, y_test = train_test_split(
    features, target, test_size=0.2, stratify=target
)
train_index = y_train[y_train == 1].index
train_data = x_train.loc[train_index]
```

```
#scaling the data using MinMaxScaler
min_max_scaler = MinMaxScaler(feature_range=(0, 1))
x_train_scaled = min_max_scaler.fit_transform(train_data.copy())
x_test_scaled = min_max_scaler.transform(x_test.copy())
```

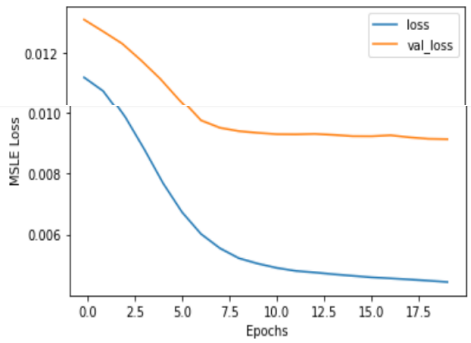
```
#creating autoencoder subclass by extending Model class from keras
class AutoEncoder(Model):
  def __init__(self, output_units, ldim=8):
    super().__init__()
    self.encoder = Sequential([
      Dense(64, activation='relu'),
      Dropout(0.1),
      Dense(32, activation='relu'),
      Dense(32, activation='relu'),
      Dropout(0.1),
      Dense(16, activation='relu'),
      Dropout(0.1),
      Dense(ldim, activation='relu')
    ])
    self.decoder = Sequential([
      Dense(16, activation='relu'),
      Dropout(0.1),
      Dense(32, activation='relu'),
      Dropout(0.1),
      Dense(64, activation='relu'),
      Dropout(0.1),
      Dense(output_units, activation='sigmoid')
    ])
  def call(self, inputs):
    encoded = self.encoder(inputs)
    decoded = self.decoder(encoded)
    return decoded
```

```
#model configuration
model = AutoEncoder(output_units=x_train_scaled.shape[1])
model.compile(loss='msle', metrics=['mse'], optimizer='adam')
epochs = 20
history = model.fit(
    x_train_scaled,
    x_train_scaled,
    epochs=epochs,
    batch_size=512,
    validation_data=(x_test_scaled, x_test_scaled)
```

```
)
```

```
Epoch 1/20
5/5 [==============================] - 7s 158ms/step - loss: 0.0112 - mse: 0.0249 - val_loss: 0.0131 - val_mse: 0.0301
Epoch 2/20
5/5 [==============================] - 0s 33ms/step - loss: 0.0107 - mse: 0.0239 - val_loss: 0.0127 - val_mse: 0.0293
Epoch 3/20
5/5 [==============================] - 0s 35ms/step - loss: 0.0098 - mse: 0.0220 - val_loss: 0.0123 - val_mse: 0.0284
Epoch 4/20
5/5 [==============================] - 0s 45ms/step - loss: 0.0088 - mse: 0.0197 - val_loss: 0.0118 - val_mse: 0.0271
Epoch 5/20
5/5 [==============================] - 0s 48ms/step - loss: 0.0077 - mse: 0.0171 - val_loss: 0.0111 - val_mse: 0.0256
Epoch 6/20
5/5 [==============================] - 0s 36ms/step - loss: 0.0067 - mse: 0.0150 - val_loss: 0.0104 - val_mse: 0.0239
Epoch 7/20
5/5 [==============================] - 0s 42ms/step - loss: 0.0060 - mse: 0.0134 - val_loss: 0.0097 - val_mse: 0.0224
Epoch 8/20
5/5 [==============================] - 0s 38ms/step - loss: 0.0055 - mse: 0.0123 - val_loss: 0.0095 - val_mse: 0.0219
Epoch 9/20
5/5 [==============================] - 0s 31ms/step - loss: 0.0052 - mse: 0.0116 - val_loss: 0.0094 - val_mse: 0.0216
```

```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.xlabel('Epochs')
plt.ylabel('MSLE Loss')
plt.legend(['loss', 'val_loss'])
plt.show()
```




```
#finding threshold for anomaly and doing predictions
def find_threshold(model, x_train_scaled):
  reconstructions = model.predict(x_train_scaled)
  reconstruction_errors = tf.keras.losses.msle(reconstructions, x_train_scaled)
  threshold = np.mean(reconstruction_errors.numpy()) \
   + np.std(reconstruction_errors.numpy())
  return threshold

def get_predictions(model, x_test_scaled, threshold):
  predictions = model.predict(x_test_scaled)
  errors = tf.keras.losses.msle(predictions, x_test_scaled)
  anomaly_mask = pd.Series(errors) > threshold
  preds = anomaly_mask.map(lambda x: 0.0 if x == True else 1.0)
  return preds

threshold = find_threshold(model, x_train_scaled)
print(f"Threshold: {threshold}")
```

```
73/73 [==============================] - 0s 3ms/step
Threshold: 0.009583446695626278
```

```
#getting accuracy score
predictions = get_predictions(model, x_test_scaled, threshold)
accuracy_score(predictions, y_test)
```

```
32/32 [==============================] - 0s 4ms/step
0.949
```