Analysis Report

The experiments are carried out using TensorFlow and Keras, which are wrappers over scikit.

Keras is a central part of the tightly-connected TensorFlow 2.0 ecosystem, covering every step of the machine learning workflow, from data management to hyperparameter training to deployment solutions.

In the beginning divided the MNIST into pairs of classes and ran it through different MLP models consisting of hidden layers [1,2,3,4]. The accuracy for each hidden layer captured then and using the best model (in our case almost all are good) took 4 hidden layers and experimented on different parameters. Similar tasks were done for Kaggle Data set too.

Processing the model using Node = [12,24,36,48,60] obtained "Accuracy" [0.9995, 0.9995,0.9991, 1.000 and 0.9991] for MNIST Data set and "MSE" for Kaggle Data set[0.1578,0.1578,0.1483,0.1741,0.1381,0.1578]

Processing the model using
Tanh, Relu, and Logistic activation functions.

[Tanh , Relu and Logistic activation Fn] obtained
'Accuracy' [0.9981, 0.9991, 0.9981] and 'MSE" for
Kaggle Data set [0.1394,0.1311,0.1338]

Processing the model using

```
Learning Rate [ 0.001, 0.010, 0.1, 1]
obtained "Accuracy' [ 0.9981, 0.9986, 0.9991,0.9986,]
and MSE" for Kaggle Data set [
0.136,0.1640,0.1421,0.1283,0.1316]
```

Processing the model using

```
Momentum [ 0.0, 0.2, 0.4, 0.9] obtained "Accuracy" [ 0.9995, 0.9986, 0.9981,0.9986] and "MSE" for Kaggle Data set [ 0.1234,0.1282,0.1456,0.1348]
```

Processing the model using

```
Early Stopping [min, max,auto] obtained "Accuracy" [ 0.9976, 0.9991,0.9986] and "MSE" for Kaggle Data set [ 0.1553,0.1673,0,1419]
```

Overall Accuracy is 100% with different inputs of the above parameters
And MSE is 0.1553.

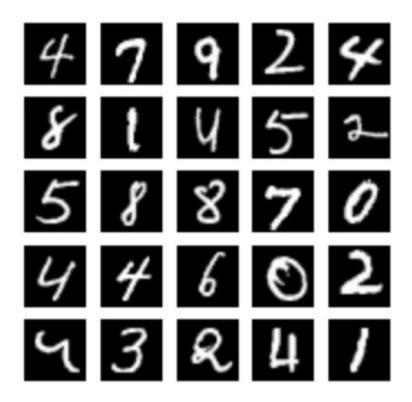
MLP processing for MNIST Analysis

The MNIST database of handwritten digits has a training set of 60,000 examples, and a test set of 10,000 examples. It is a subset of a larger set available from NIST. The digits have been size-normalized and centered in a fixed-size image.

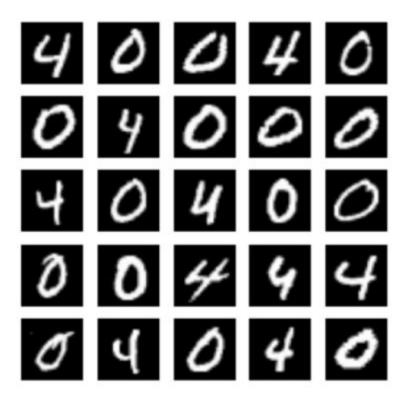
MLP Analytic Results:

Executed MLP on MNIST dataset using different classes using layers 1,2,3 and 4 as shown below

```
plt.figure(figsize=(5,5))
for i in range(len(indexes)):
   plt.subplot(5, 5, i + 1)
   image = images[i]
   plt.imshow(image, cmap='gray')
   plt.axis('off')
```



Selected 2 classes [0,4] as required for experimenting:



Train an MLP each with 1, 2, 3 and 4 hidden layers using Backpropagation.

Defined the MLP model Using Two Hidden Layer

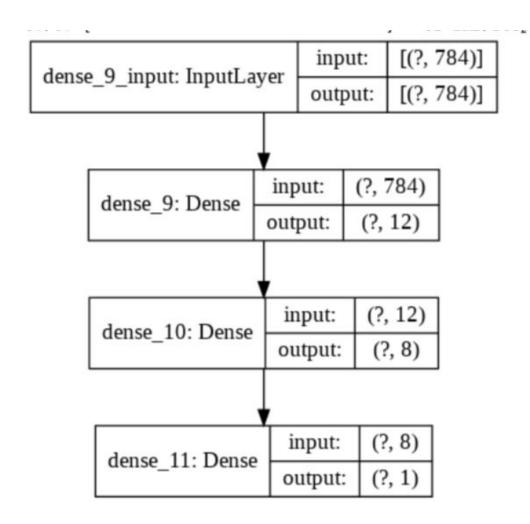
```
model = Sequential()
model.add(Dense(12, input_shape=(784,), activation='relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

Results

2.4.0 Model: "sequential 4"

Layer (type)	Output Shape	Param #
dense_9 (Dense)	(None, 12)	9420
dense_10 (Dense)	(None, 8)	104
dense_11 (Dense)	(None, 1)	9

Plotted the Layer Diagram showing Input/Output -



```
model.compile(optimizer='sgd', loss='binary_crossentropy',
metrics=['binary_accuracy'])
model.fit(
    x=x_train_final,
    y=y_train_new,
    shuffle=True,
    epochs=5,
    batch_size=16
)
```

Output:

Epoch 1/5

Accuracy -----> 0.9995

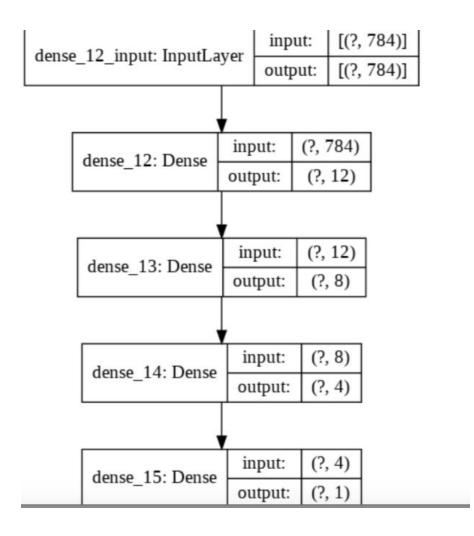
Using Three Hidden Layers

```
model = Sequential()
model.add(Dense(12, input_shape=(784,), activation='relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(4, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

2.4.0 Model: "sequential 5"

Layer (type)	Output Shape	Param #
dense_12 (Dense)	(None, 12)	9420
dense_13 (Dense)	(None, 8)	104
dense_14 (Dense)	(None, 4)	36
dense_15 (Dense)	(None, 1)	5

Total params: 9,565
Trainable params: 9,565
Non-trainable params: 0



67/67 [==============] - 0s 2ms/step - loss: 0.0020 - binary accuracy: 0.9999 or 1 or 100

Accuracy ---->0.9999

Using 1 Hidden Layer

```
model = Sequential()

model.add(Dense(1, input_shape=(784,)))

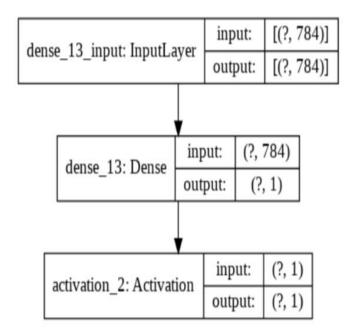
model.summary()
plot_model(model, to_file='mlp-mnist.png', show_shapes=True)
```

2.4.0 Model: "sequential_3"

Layer (type)	Output Shape	Param #
dense_8 (Dense)	(None, 1)	785
activation_1 (Activation)	(None, 1)	0

Total params: 785
Trainable params: 785
Non-trainable params: 0

binary_accuracy: 0.9976



Accuracy ---->0.9995

From the above experiments, all the models [1,2,3,4] have very good Accuracy almost 100.

So for further experiments, 4 hidden layers are considered.

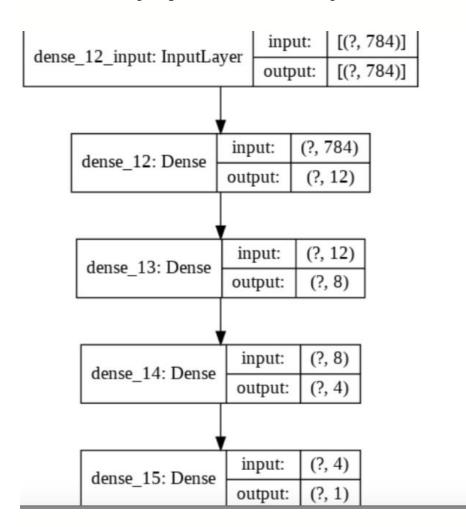
Experimented to determine the Best Batch and Epoch

The batch size is the number of patterns shown to the network before the weights are updated. It is also an optimization in the training of the network, defining how many patterns to read at a time and keep in memory.

The number of epochs is the number of times that the entire training dataset is shown to the network during training

4 Hidden Layer is performed better than other hidden layers however the difference is not very MUCH. It is really close.

For fine tuning experiment considering the 4 Hidden Layer



Experiment with Number of nodes in the hidden layers.

Processing MLP by changing the number of Nodes passed to the Model.

Considering following no of nodes for experiment

Node = [12,24,36,48,60]

Processing - Node changes 12

Model: "sequential 1"

binary accuracy: 0.9989

Layer (type)	Output	Shape	Param	 l #
dense_4 (Dense)	(None,	12)	9420	:====
dense_5 (Dense)	(None,	8)	104	
dense_6 (Dense)	(None,	4)	36	
dense_7 (Dense)	(None,		5	
Total params: 9,565 Trainable params: 9,565 Non-trainable params: 0				
Epoch 1/5 792/792 [====================================		====] - 2s :	2ms/step - los	ss: 0.1548 -
Epoch 2/5 792/792 [====================================		====] - 2s :	2ms/step - los	s: 0.0086 -
Epoch 3/5 792/792 [====================================	======	====] - 2s	2ms/step - los	s: 0.0055 -
Epoch 4/5 792/792 [====================================		====] - 2s :	2ms/step - los	s: 0.0044 -
Epoch 5/5 792/792 [====================================	======	====] - 2s	2ms/step - los	s: 0.0038 -

```
67/67 [=========] - 0s 2ms/step -
```

loss: 0.0017 - binary_accuracy: 0.9995

2.4.0 Processing - Node changes 24

Model: "sequential 2"

Layer (type)	Output	Shape	Param #
dense_8 (Dense)	(None,	24)	18840
dense_9 (Dense)	(None,	20)	500
dense_10 (Dense)	(None,	16)	336
dense_11 (Dense)	(None,	1)	17
Total params: 19,693 Trainable params: 19,693 Non-trainable params: 0			
Epoch 1/5 792/792 [====================================	======	===] - 2s 2ms/step	- loss: 0.0605 -
792/792 [====================================		====] - 2s 2ms/step	- loss: 0.0070 -
Epoch 3/5 792/792 [====================================	======	===] - 2s 2ms/step	- loss: 0.0050 -

Epoch 4/5

792/792 [===========] - 2s 2ms/step - loss: 0.0041 -

binary accuracy: 0.9987

Epoch 5/5

792/792 [============] - 2s 2ms/step - loss: 0.0035 -

binary_accuracy: 0.9990

binary_accuracy: 0.9995

2.4.0

Processing - Node changes 36

Model: "sequential_3"

Layer (type)	Output	Shape	Param #
dense_12 (Dense)	(None,	36)	28260
dense_13 (Dense)	(None,	32)	1184
dense_14 (Dense)	(None,	28)	924
dense_15 (Dense)	(None,		 29 =======
Total params: 30,397 Trainable params: 30,397 Non-trainable params: 0			
Epoch 1/5 792/792 [====================================	:=====:	===] - 2s 2ms/step	- loss: 0.0988 -
Epoch 2/5 792/792 [====================================	:=====	===] - 2s 2ms/step	- loss: 0.0091 -
Epoch 3/5 792/792 [====================================	:=====:	===] - 2s 2ms/step	- loss: 0.0064 -
792/792 [====================================	:=====:	===] - 2s 2ms/step	- loss: 0.0052 -
Epoch 5/5 792/792 [====================================			
67/67 [====================================		==] - 0s 2ms/step -	loss: 0.0026 -

Processing - Node changes 48

Model: "sequential_4"

Layer (type)	Output Shape	Param #
dense_16 (Dense)	(None, 48)	37680
dense_17 (Dense)	(None, 44)	2156
dense_18 (Dense)	(None, 40)	1800

dense_19 (Dense)	(None, 1)	41	
Total params: 41,677 Trainable params: 41,677 Non-trainable params: 0			==
Epoch 1/5 792/792 [====================================] - 2s	2ms/step - loss:	0.1060 -
Epoch 2/5 792/792 [====================================	=====] - 2s	3ms/step - loss:	0.0079 -
Epoch 3/5 792/792 [====================================	=====] - 2s	2ms/step - loss:	0.0057 -
Epoch 4/5 792/792 [====================================	=====] - 2s	2ms/step - loss:	0.0047 -
Epoch 5/5 792/792 [====================================	=====] - 2s	2ms/step - loss:	0.0040 -

67/67 [==============] - 0s 2ms/step - loss: 0.0018 - binary_accuracy: 1.0000

2.4.0

Processing - Node changes 60

Model: "sequential_5"

Layer (type)	Output Shape	Param #
dense_20 (Dense)	(None, 60)	47100
dense_21 (Dense)	(None, 56)	3416
dense_22 (Dense)	(None, 52)	2964
dense_23 (Dense)	(None, 1)	53

Total params: 53,533
Trainable params: 53,533
Non-trainable params: 0

```
Epoch 1/5
binary accuracy: 0.9925
Epoch 2/5
792/792 [============ ] - 2s 3ms/step - loss: 0.0075 -
binary accuracy: 0.9979
Epoch 3/5
binary accuracy: 0.9987
Epoch 4/5
binary accuracy: 0.9989
Epoch 5/5
binary accuracy: 0.9991
binary accuracy: 0.9991
```

Processing the model using Tanh, Relu, and Logistic activation functions.

The activation function controls the non-linearity of individual neurons and when to fire.

Processing - activation function changes ---> "tanh"

Model: "sequential 6"

Layer (type)	Outpu	t Shape	Param #
dense_24 (De	nse) (None	, 16)	12560

dense_25 (Dense)	(None, 12)	204	
dense_26 (Dense)	(None, 8)	104	
dense_27 (Dense)	(None, 1)	9	
Total params: 12,877 Trainable params: 12,877 Non-trainable params: 0			
Epoch 1/5 792/792 [====================================] - 2s 2m	s/step - loss: 0.0	238 -
792/792 [====================================] - 2s 2m	s/step - loss: 0.0	102 -
792/792 [====================================	======] - 2s 2m	s/step - loss: 0.0	094 -
792/792 [====================================	======] - 2s 2m	s/step - loss: 0.0	088 -
792/792 [====================================			
binary_accuracy: 0.9981		scep - 1055. 0.010	.

Processing - activation function changes 'relu'

Model: "sequential_7"

Layer (ty	/pe)	Output	Shape	Param #
dense_28	(Dense)	(None,	16)	12560
dense_29	(Dense)	(None,	12)	204
dense_30	(Dense)	(None,	8)	104
dense_31	(Dense)	(None,	1)	9

Total params: 12,877

Trainable params: 12,877 Non-trainable params: 0

Epoch 1/5 binary accuracy: 0.9931 Epoch 2/5 binary accuracy: 0.9987 Epoch 3/5 binary accuracy: 0.9984 Epoch 4/5 binary accuracy: 0.9991 Epoch 5/5 binary accuracy: 0.9988 binary accuracy: 0.9991 2.4.0

Processing - activation function changes 'sigmoid'

Model: "sequential 8"

Layer (type)	Output Shape	Param #
dense_32 (Dense)	(None, 16)	12560
dense_33 (Dense)	(None, 12)	204
dense_34 (Dense)	(None, 8)	104
dense_35 (Dense)	(None, 1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

```
binary accuracy: 0.5229
Epoch 2/5
792/792 [============ ] - 2s 2ms/step - loss: 0.6662 -
binary accuracy: 0.5552
Epoch 3/5
binary accuracy: 0.7484
Epoch 4/5
792/792 [============ ] - 2s 2ms/step - loss: 0.4882 -
binary accuracy: 0.9843
Epoch 5/5
binary accuracy: 0.9958
binary accuracy: 0.9981
```

Activation function achieved the best results with an accuracy of about 99.9 [almost 100%]

Processing Hyperparameters -

#Hyperparameters: Momentum term, Early stopping, and Learning Rate

Learning rate controls how much to update the weight at the end of each batch and the momentum controls how much to let the previous update influence the current weight update.

Will try a suite of small standard learning rates as given below with fixed batch size and epochs.

learning rate list = [0.001, 0.010, 0.1, 1]

Processing - learning rate changes 0.001

Model: "sequential_19"

Layer (type)	Output Shape	Param #
dense_76 (Dense)	(None, 16)	12560
dense_77 (Dense)	(None, 12)	204
dense_78 (Dense)	(None, 8)	104
dense_79 (Dense)	(None, 1)	9

Total params: 12,877 Trainable params: 12,877 Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f0853d88c18>

Epoch 1/5

binary accuracy: 0.9558

Epoch 2/5

binary accuracy: 0.9987

Epoch 3/5

binary accuracy: 0.9993

Epoch 4/5

binary_accuracy: 0.9994

Epoch 5/5

binary accuracy: 0.9995

binary accuracy: 0.9981

Processing - learning rate changes 0.01

Model: "sequential_20"

dense_80	(Dense)	(None, 16)	12560		
dense_81	(Dense)	(None, 12)	204		
dense_82	(Dense)	(None, 8)	104		
dense_83	(Dense)	(None, 1)	9		
Total params: 12,877 Trainable params: 12,877					

ainable params: 12,8 Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f085c22bdd8>

Epoch 1/5

binary accuracy: 0.9582

Epoch 2/5

binary accuracy: 0.9988

Epoch 3/5

binary accuracy: 0.9995

Epoch 4/5

binary accuracy: 0.9994

Epoch 5/5

binary accuracy: 0.9993

binary accuracy: 0.9986

2.4.0

Processing - learning rate changes 0.1

Model: "sequential 21"

Layer (ty	pe)	Output	Shape	Param #
dense_84	(Dense)	(None,	16)	12560
dense_85	(Dense)	(None,	12)	204
dense_86	(Dense)	(None,	8)	104
dense_87	(Dense)	(None,	1)	9

Total params: 12,877

Trainable params: 12,877 Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at
0x7f0851a3dac8>

Epoch 1/5

binary accuracy: 0.9721

Epoch 2/5

binary accuracy: 0.9988

Epoch 3/5

792/792 [============] - 2s 2ms/step - loss: 0.0166 -

binary accuracy: 0.9992

Epoch 4/5

binary_accuracy: 0.9994

Epoch 5/5

binary accuracy: 0.9993

binary_accuracy: 0.9991

2.4.0

Processing - learning rate changes 1

Model: "sequential 22"

Layer (ty		Output		Param #
dense_88		(None,	16)	12560
dense_89	(Dense)	(None,	12)	204
dense_90	(Dense)	(None,	8)	104
dense_91	(Dense)	(None,	1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f0852b86d30>

Epoch 1/5

binary_accuracy: 0.9464

Epoch 2/5

Processing momentum:

Experiment with different momentum values [0.0,0.2,0.4,0.9], keeping others parameters fixed.

$$momentum_list = [0.0, 0.2, 0.4, 0.9]$$

2.4.0 Processing - momentum changes 0.0

Model: "sequential 42"

Layer (typ	oe)	Output	Shape	Param #
dense_168	(Dense)	(None,	16)	12560
dense_169	(Dense)	(None,	12)	204
dense_170	(Dense)	(None,	8)	104
dense_171	(Dense)	(None,	1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f086e0a75c0>

binary accuracy: 0.9799 Epoch 2/5 binary accuracy: 0.9987 Epoch 3/5 binary accuracy: 0.9991 Epoch 4/5 792/792 [============] - 2s 2ms/step - loss: 0.0019 binary accuracy: 0.9996 Epoch 5/5 binary accuracy: 0.9990 binary accuracy: 0.9995 2.4.0

Processing - momentum changes 0.2

Model: "sequential 43"

Layer (type)	Output Shape	Param #
dense_172 (Dense)	(None, 16)	12560
dense_173 (Dense)	(None, 12)	204
dense_174 (Dense)	(None, 8)	104
dense_175 (Dense)	(None, 1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f0854f74a20>

binary_accuracy: 0.9988

Epoch 3/5

Processing - momentum changes 0.4

Model: "sequential_44"

Layer (type)	Output Shape	Param #
dense_176 (Dense)	(None, 16)	12560
dense_177 (Dense)	(None, 12)	204
dense_178 (Dense)	(None, 8)	104
dense_179 (Dense)	(None, 1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f08fdda52b0>

Epoch 5/5

binary accuracy: 0.9989

binary_accuracy: 0.9981

2.4.0

Processing - momentum changes 0.9

Model: "sequential_45"

Layer (type)	Output Shape	Param #
dense_180 (Dense)	(None, 16)	12560
dense_181 (Dense)	(None, 12)	204
dense_182 (Dense)	(None, 8)	104
dense_183 (Dense)	(None, 1)	9

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at 0x7f0852c22d68>

```
Epoch 1/5
```

792/792 [============] - 2s 2ms/step - loss: 0.0374 -

binary accuracy: 0.9848

Epoch 2/5

binary accuracy: 0.9968

Epoch 3/5

binary_accuracy: 0.9969

Epoch 4/5

binary_accuracy: 0.9981

Epoch 5/5

binary accuracy: 0.9986

Accuracy is almost [100% for different values of momentum]

Experimentation with early_stopping_list..

Early Stopping is used to Halt the Training of Neural Networks At the Right Time..

A problem with training neural networks is in the choice of the number of training epochs to use.

Too many epochs can lead to overfitting of the training dataset, whereas too few may result in an underfit model. Early stopping is a method that allows you to specify an arbitrary large number of training epochs and stop training once the model performance stops improving on a hold out validation dataset.

```
early_stopping_list = [ 'min', 'max', 'auto']

for e_s in early_stopping_list:

    callback = k.callbacks.EarlyStopping(monitor='accuracy',
min_delta=0.0001,patience=1)

2.4.0

Processing - early stopping changes min
Model: "sequential_8"

Layer (type) Output Shape Param #
```

```
dense_32 (Dense)
                    (None, 16)
                                      12560
dense 33 (Dense)
                    (None, 12)
                                      204
dense 34 (Dense)
                    (None, 8)
                                      104
dense 35 (Dense)
                    (None, 1)
Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0
model <tensorflow.python.keras.engine.sequential.Sequential object at
0x7f284e291ac8>
Epoch 1/100
accuracy: 0.9756 - val loss: 0.0048 - val accuracy: 0.9987
Epoch 2/100
accuracy: 0.9973 - val loss: 0.0035 - val accuracy: 0.9989
Epoch 3/100
accuracy: 0.9982 - val loss: 0.0033 - val accuracy: 0.9995
Epoch 4/100
accuracy: 0.9982 - val loss: 0.0128 - val accuracy: 0.9955
No of epochs ran 4
accuracy: 0.9976
2.4.0
Processing - early stopping changes max
Model: "sequential 9"
Layer (type)
                    Output Shape
                                      Param #
dense 36 (Dense)
                    (None, 16)
                                      12560
dense 37 (Dense)
                    (None, 12)
                                      204
```

Total params: 12,877
Trainable params: 12,877
Non-trainable params: 0

model <tensorflow.python.keras.engine.sequential.Sequential object at
0x7f284e0d60b8>

Epoch 1/100

accuracy: 0.9748 - val_loss: 0.0048 - val_accuracy: 0.9987

Epoch 2/100

accuracy: 0.9984 - val_loss: 0.0045 - val_accuracy: 0.9992

Epoch 3/100

accuracy: 0.9990 - val_loss: 0.0029 - val_accuracy: 0.9995

Epoch 4/100

accuracy: 0.9984 - val_loss: 0.0029 - val_accuracy: 0.9995

No of epochs ran 4

accuracy: 0.9991

2.4.0

Processing - early stopping changes auto

Model: "sequential 10"

Layer (type)	Output Shape	Param #
dense_40 (Dense)	(None, 16)	12560
dense_41 (Dense)	(None, 12)	204
dense_42 (Dense)	(None, 8)	104
dense_43 (Dense)	(None, 1)	9

Total params: 12,877

Trainable params: 12,877 Non-trainable params: 0

```
model <tensorflow.python.keras.engine.sequential.Sequential object at
0x7f2842742e10>
Epoch 1/100
accuracy: 0.9649 - val_loss: 0.0044 - val_accuracy: 0.9984
Epoch 2/100
accuracy: 0.9966 - val loss: 0.0040 - val accuracy: 0.9987
Epoch 3/100
accuracy: 0.9984 - val loss: 0.0025 - val accuracy: 0.9997
Epoch 4/100
accuracy: 0.9992 - val loss: 0.0035 - val accuracy: 0.9992
Epoch 5/100
accuracy: 0.9985 - val_loss: 0.0028 - val_accuracy: 0.9997
No of epochs ran 5
67/67 [============= ] - 0s 2ms/step - loss:0.0046 -
accuracy: 0.9986
```

Kaggle Analysis Report

Kaggle Analysis:

For IBM Dataset to determine the Attrition in the company using MLP

The Features consists of both Numerical and Categorical Data as shown below: Also attached the HR.CSV sheet for reference.

```
['Age', 'DailyRate', 'DistanceFromHome', 'Education', 'EmployeeCount', 'EmployeeNumber', 'EnvironmentSatisfaction', 'HourlyRate', 'JobInvolvement', 'JobLevel', 'JobSatisfaction', 'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked', 'PercentSalaryHike', 'PerformanceRating', 'RelationshipSatisfaction', 'StandardHours', 'StockOptionLevel', 'TotalWorkingYears', 'TrainingTimesLastYear', 'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole', 'YearsSinceLastPromotion', 'YearsWithCurrManager', 'n_Gender', 'n_JobRole', 'n_BusinessTravel', 'n_Department', 'n_EducationField'], dtype='object')
```

In [11]:

	Age	Attrition	BusinessTravel	DailyRate	Department	DistanceFromHome	Education	EducationField	EmployeeCount	EmployeeNumber	Env
0	41	Yes	Travel_Rarely	1102	Sales	1	2	Life Sciences	1	1	
1	49	No	Travel_Frequently	279	Research & Development	8	1	Life Sciences	1	2	
2	37	Yes	Travel_Rarely	1373	Research & Development	2	2	Other	1	4	
3	33	No	Travel_Frequently	1392	Research & Development	3	4	Life Sciences	1	5	
4	27	No	Travel_Rarely	591	Research & Development	2	1	Medical	1	7	

Identified best model among layer [1,2,3,4] hidden layer

Experimented - Using 3 hidden layers

```
model = Sequential()
```

```
model.add(Dense(16, input shape=(31,), activation='relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(4, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
#---
```

Layer (ty	rpe)	Output	Shape	Param #
dense_28	(Dense)	(None,	12)	384
dense_29	(Dense)	(None,	8)	104
dense_30	(Dense)	(None,	4)	36
dense_31	(Dense)	(None,	1)	5

Total params: 529

Trainable params: 529 Non-trainable params: 0

```
Epoch 1/100
```

2/2 - 0s - loss: 0.1578 - mse: 0.1578 - mae: 0.1578 - mape: 15.7823

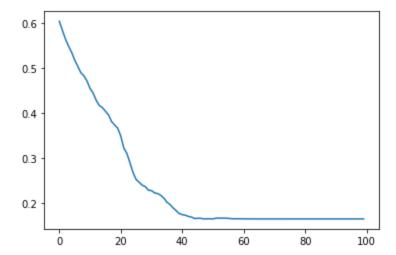
Epoch 2/100

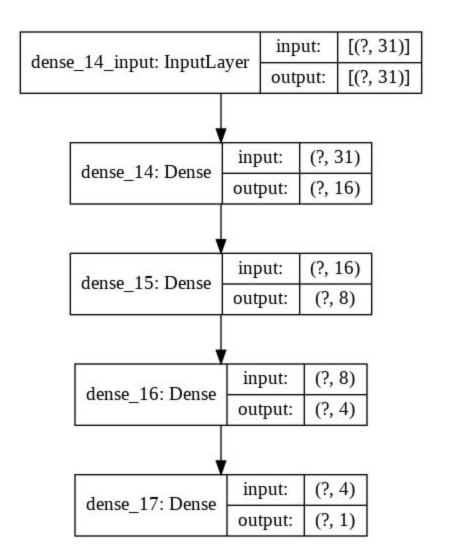
2/2 - 0s - loss: 0.1578 - mse: 0.1578 - mae: 0.1578 - mape: 15.7823

Epoch 3/100

2/2 - 0s - loss: 0.1578 - mse: 0.1578 - mae: 0.1578 - mape: 15.7823

Epoch 4/100





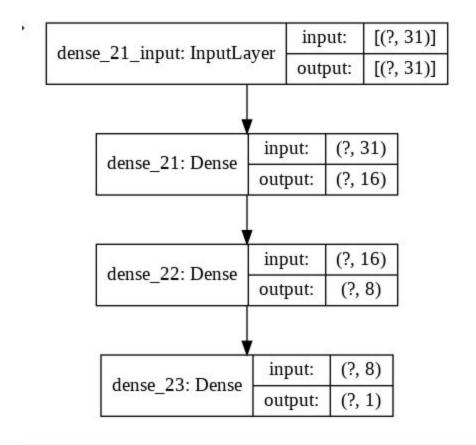
MSE 0.1578

Experimenting with 2 Hidden Layers

```
model = Sequential()
model.add(Dense(16, input_shape=(31,), activation='relu'))
model.add(Dense(8, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
```

```
Epoch 88/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 89/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 90/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 91/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 92/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 93/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 94/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 95/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 96/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8711
Epoch 97/100
```

MSE: 0.1687



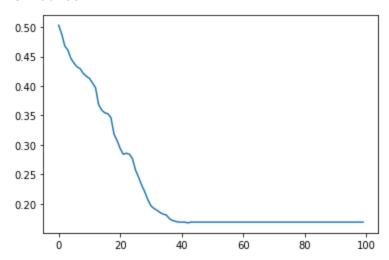
Experimenting with 1 Hidden Layer

```
model = Sequential()
  model.add(Dense(16, input_shape=(31,), activation='relu'))
  model.add(Dense(1, activation='sigmoid'))
#---

Epoch 29/100
1/1 - 0s - loss: 0.2202 - mse: 0.2202 - mae: 0.2211 - mape: 89027704.0000
Epoch 30/100
1/1 - 0s - loss: 0.2070 - mse: 0.2070 - mae: 0.2073 - mape: 74012832.0000
Epoch 31/100
1/1 - 0s - loss: 0.1965 - mse: 0.1965 - mae: 0.1971 - mape: 56738072.0000
Epoch 32/100
1/1 - 0s - loss: 0.1919 - mse: 0.1919 - mae: 0.1921 - mape: 42445088.0000
Epoch 33/100
```

```
1/1 - 0s - loss: 0.1889 - mse: 0.1889 - mae: 0.1891 - mape: 35438988.0000
Epoch 34/100
1/1 - 0s - loss: 0.1850 - mse: 0.1850 - mae: 0.1850 - mape: 28571500.0000
Epoch 35/100
1/1 - 0s - loss: 0.1823 - mse: 0.1823 - mae: 0.1823 - mape: 23129268.0000
Epoch 36/100
1/1 - 0s - loss: 0.1808 - mse: 0.1808 - mae: 0.1809 - mape: 17664676.0000
Epoch 37/100
1/1 - 0s - loss: 0.1747 - mse: 0.1747 - mae: 0.1750 - mape: 10402883.0000
Epoch 38/100
1/1 - 0s - loss: 0.1714 - mse: 0.1714 - mae: 0.1714 - mape: 5442194.0000
Epoch 39/100
1/1 - 0s - loss: 0.1701 - mse: 0.1701 - mae: 0.1701 - mape: 2721105.2500
Epoch 40/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 1360741.1250
Epoch 41/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 1360560.8750
Epoch 42/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 1360560.3750
Epoch 43/100
1/1 - 0s - loss: 0.1673 - mse: 0.1673 - mae: 0.1674 - mape: 16.7600
Epoch 44/100
1/1 - 0s - loss: 0.1687 - mse: 0.1687 - mae: 0.1687 - mape: 16.8707
Epoch 45/100
```

MSE 0.1687



From the experiment above, the MSE error is almost same for all model based on Hidden Layer 1,2,3 so choosing 3 layer for experimenting with parameters:

#MLP Processing by changing the NODEs in the Model

```
Choosed following Model for processing - 3 Hidden Layer
   Node = 16
    model = Sequential()
    model.add(Dense(node, input shape=(31,), activation='relu'))
    model.add(Dense(node-4, activation='relu'))
    model.add(Dense(node-8, activation='relu'))
    model.add(Dense(1, activation='sigmoid'))
Model: "sequential 67"
Layer (type)
                          Output Shape
                                                 Param #
______
dense 308 (Dense)
                          (None, 12)
                                                 384
dense 309 (Dense)
                          (None, 8)
                                                 104
dense 310 (Dense)
                          (None, 4)
                                                 36
dense 311 (Dense)
                          (None, 1)
                                                 5
______
Total params: 529
Trainable params: 529
Non-trainable params: 0
Epoch 1/100
2/2 - 0s - loss: 0.5510 - mse: 0.5510 - mae: 0.5510 - mape: 459883104.0000
Epoch 2/100
2/2 - 0s - loss: 0.5428 - mse: 0.5428 - mae: 0.5429 - mape: 447681984.0000
Epoch 3/100
2/2 - 0s - loss: 0.5417 - mse: 0.5417 - mae: 0.5422 - mape: 448318752.0000
Epoch 4/100
2/2 - 0s - loss: 0.5294 - mse: 0.5294 - mae: 0.5298 - mape: 433188416.0000
```

Epoch 5/100

2/2 - 0s - loss: 0.5245 - mse: 0.5245 - mae: 0.5248 - mape: 425491552.0000

Epoch 6/100

2/2 - 0s - loss: 0.5235 - mse: 0.5235 - mae: 0.5237 - mape: 418908256.0000

Epoch 7/100

2/2 - 0s - loss: 0.5131 - mse: 0.5131 - mae: 0.5135 - mape: 407387104.0000

Epoch 96/100

2/2 - 0s - loss: 0.2122 - mse: 0.2122 - mae: 0.2122 - mape: 54421992.0000

Epoch 97/100

2/2 - 0s - loss: 0.2063 - mse: 0.2063 - mae: 0.2066 - mape: 48770628.0000

Epoch 98/100

2/2 - 0s - loss: 0.2041 - mse: 0.2041 - mae: 0.2041 - mape: 46258520.0000

Epoch 99/100

2/2 - 0s - loss: 0.2041 - mse: 0.2041 - mae: 0.2041 - mape: 44897164.0000

Epoch 100/100

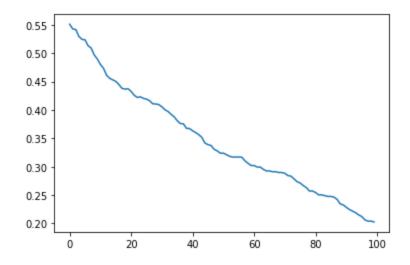
MSE : 0.2041

Processing - Node changes 60

Model: "sequential_74"

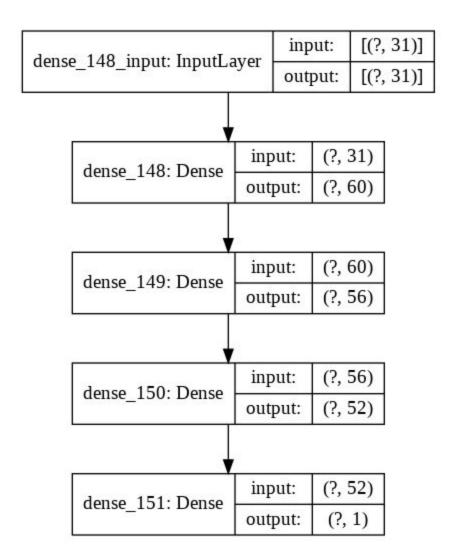
Layer (type)	Output Shape	Param #	
dense_336 (Dense)	(None, 60)	1920	
dense_337 (Dense)	(None, 56)	3416	
dense_338 (Dense)	(None, 52)	2964	
dense_339 (Dense)	(None, 1)	53	

Total params: 8,353 Trainable params: 8,353 Non-trainable params: 0



```
Epoch 89/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 90/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 91/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 92/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 93/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 94/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 95/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 96/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 97/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 98/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 99/100
2/2 - 0s - loss: 0.1660 - mse: 0.1660 - mae: 0.1660 - mape: 16.5986
Epoch 100/100
```

MSE: 0.1660



Tanh, Relu, and Logistic activation functions.

```
fn_list = [ "tanh", "relu", "sigmoid"]

print(f" Processing - activation function changes { fn }")
    node = 16 # No. of nodes - fixed
    model.add(Dense(node , input_shape=(31,), activation=fn))
    model.add(Dense(node -4, activation=fn))
    model.add(Dense(node - 8, activation=fn))
```

Processing - activation function changes tanh Model: "sequential_55"

Layer (type) Out	out Shape Param #	_
dense_236 (Dense) (No.	ne, 16) 512	=
dense_237 (Dense) (No.	ne, 12) 204	_
dense_238 (Dense) (No.	ne, 8) 104	_
dense_239 (Dense) (No:	ne, 1) 9	_
dense_240 (Dense) (No.	ne, 16) 32	_
dense_241 (Dense) (No.	ne, 12) 204	_
dense_242 (Dense) (No.	ne, 8) 104	_
dense_243 (Dense) (No.	ne, 1) 9	=
Epoch 1/100		
2/2 - 0s - loss: 0.2485 - mse: 0 Epoch 2/100	.2485 - mae: 0.4984 - mape: 4149	11616.0000
2/2 - 0s - loss: 0.2175 - mse: 0 Epoch 3/100	.2175 - mae: 0.4645 - mape: 3727	13312.0000
2/2 - 0s - loss: 0.1901 - mse: 0 Epoch 4/100	.1901 - mae: 0.4286 - mape: 3281	04832.0000
2/2 - 0s - loss: 0.1684 - mse: 0 Epoch 5/100	.1684 - mae: 0.3921 - mape: 2821	68896.0000
2/2 - 0s - loss: 0.1519 - mse: 0 Epoch 6/100	.1519 - mae: 0.3544 - mape: 2354	23056.0000
2/2 - 0s - loss: 0.1424 - mse: 0 Epoch 7/100	.1424 - mae: 0.3183 - mape: 1901	64912.0000
2/2 - 0s - loss: 0.1389 - mse: 0 Epoch 8/100	.1389 - mae: 0.2849 - mape: 1483	23472.0000
2/2 - 0s - loss: 0.1393 - mse: 0 Epoch 9/100	.1393 - mae: 0.2562 - mape: 1126	90400.0000
2/2 - 0s - loss: 0.1424 - mse: 0 Epoch 10/100	.1424 - mae: 0.2350 - mape: 8617	2336.0000

Processing - activation function changes relu

Model: "sequential 55"

Layer (typ	pe)	Output	Shape	Param #
dense_236	(Dense)	(None,	16)	512
dense_237	(Dense)	(None,	12)	204
dense_238	(Dense)	(None,	8)	104
dense_239	(Dense)	(None,	1)	9
dense_240	(Dense)	(None,	16)	32
dense_241	(Dense)	(None,	12)	204
dense_242	(Dense)	(None,	8)	104
dense_243	(Dense)	(None,	1)	9
dense_244	(Dense)	(None,	16)	32
dense_245	(Dense)	(None,	12)	204
dense_246	(Dense)	(None,	8)	104
dense_247	(Dense)	(None,	1)	9
dense_248	(Dense)	(None,	16)	32
dense_249	(Dense)	(None,	12)	204
dense_250	(Dense)	(None,	8)	104
dense_251	(Dense)	(None,	1)	9

Total params: 1,876
Trainable params: 1,876
Non-trainable params: 0

Epoch 1/100

2/2 - 0s - loss: 0.1638 - mse: 0.1638 - mae: 0.1743 - mape: 8790954.0000

Epoch 2/100

2/2 - 0s - loss: 0.1601 - mse: 0.1601 - mae: 0.1826 - mape: 19203358.0000

```
2/2 - 0s - loss: 0.1565 - mse: 0.1565 - mae: 0.1914 - mape: 30119582.0000
Epoch 4/100
2/2 - 0s - loss: 0.1532 - mse: 0.1532 - mae: 0.2010 - mape: 42025232.0000
Epoch 5/100
2/2 - 0s - loss: 0.1498 - mse: 0.1498 - mae: 0.2111 - mape: 54631900.0000
Epoch 6/100
2/2 - 0s - loss: 0.1468 - mse: 0.1468 - mae: 0.2219 - mape: 68199400.0000
Epoch 7/100
2/2 - 0s - loss: 0.1440 - mse: 0.1440 - mae: 0.2333 - mape: 82511360.0000
Epoch 8/100
2/2 - 0s - loss: 0.1416 - mse: 0.1416 - mae: 0.2450 - mape: 97421848.0000
Epoch 9/100
2/2 - 0s - loss: 0.1402 - mse: 0.1402 - mae: 0.2575 - mape: 112953616.0000
Epoch 10/100
2/2 - 0s - loss: 0.1396 - mse: 0.1396 - mae: 0.2702 - mape: 128649344.0000
Epoch 11/100
2/2 - 0s - loss: 0.1394 - mse: 0.1394 - mae: 0.2815 - mape: 142849216.0000
Epoch 12/100
2/2 - 0s - loss: 0.1394 - mse: 0.1394 - mae: 0.2760 - mape: 135948880.0000
Epoch 99/100
2/2 - 0s - loss: 0.1394 - mse: 0.1394 - mae: 0.2750 - mape: 134735440.0000
Epoch 100/100
2/2 - 0s - loss: 0.1394 - mse: 0.1394 - mae: 0.2742 - mape: 133687088.0000
```

MSE - 0.1394

Epoch 3/100

Processing - activation function changes sigmoid

Model: "sequential 55"

Layer (typ	pe)	Output	Shape	Param #
=======		======		
dense_236	(Dense)	(None,	16)	512
dense_237	(Dense)	(None,	12)	204
dense_238	(Dense)	(None,	8)	104
dense_239	(Dense)	(None,	1)	9
dense_240	(Dense)	(None,	16)	32

dense_241	(Dense)	(None,	12)	204
dense_242	(Dense)	(None,	8)	104
dense_243	(Dense)	(None,	1)	9
dense_244	(Dense)	(None,	16)	32
dense_245	(Dense)	(None,	12)	204
dense_246	(Dense)	(None,	8)	104
dense_247	(Dense)	(None,	1)	9
dense_248	(Dense)	(None,	16)	32
dense_249	(Dense)	(None,	12)	204
dense_250	(Dense)	(None,	8)	104
dense_251	(Dense)	(None,	1)	9
dense_252	(Dense)	(None,	16)	32
dense_253	(Dense)	(None,	12)	204
dense_254	(Dense)	(None,	8)	104
dense_255	(Dense)	(None,	1)	9
dense_256	(Dense)	(None,	16)	32
dense_257	(Dense)	(None,	12)	204
dense_258	(Dense)	(None,	8)	104
dense_259	(Dense)	(None,	1)	9

Total params: 2,574
Trainable params: 2,574
Non-trainable params: 0

Epoch 1/100

2/2 - 0s - loss: 0.4282 - mse: 0.4282 - mae: 0.6371 - mape: 588955328.0000

Epoch 2/100

2/2 - 0s - loss: 0.4227 - mse: 0.4227 - mae: 0.6337 - mape: 584730496.0000

Epoch 3/100

```
2/2 - 0s - loss: 0.4174 - mse: 0.4174 - mae: 0.6302 - mape: 580507648.0000
Epoch 4/100
2/2 - 0s - loss: 0.4120 - mse: 0.4120 - mae: 0.6268 - mape: 576278528.0000
Epoch 5/100
2/2 - 0s - loss: 0.4067 - mse: 0.4067 - mae: 0.6234 - mape: 572015488.0000
Epoch 6/100
2/2 - 0s - loss: 0.4014 - mse: 0.4014 - mae: 0.6199 - mape: 567723840.0000
Epoch 7/100
2/2 - 0s - loss: 0.3962 - mse: 0.3962 - mae: 0.6165 - mape: 563444800.0000
Epoch 8/100
2/2 - 0s - loss: 0.3910 - mse: 0.3910 - mae: 0.6130 - mape: 559137216.0000
Epoch 9/100
2/2 - 0s - loss: 0.3858 - mse: 0.3858 - mae: 0.6095 - mape: 554819840.0000
Epoch 10/100
2/2 - 0s - loss: 0.3806 - mse: 0.3806 - mae: 0.6060 - mape: 550477184.0000
Epoch 11/100
2/2 - 0s - loss: 0.3756 - mse: 0.3756 - mae: 0.6026 - mape: 546187648.0000
Epoch 12/100
2/2 - 0s - loss: 0.3705 - mse: 0.3705 - mae: 0.5991 - mape: 541846016.0000
Epoch 13/100
2/2 - 0s - loss: 0.3656 - mse: 0.3656 - mae: 0.5956 - mape: 537533312.0000
Epoch 14/100
2/2 - 0s - loss: 0.3607 - mse: 0.3607 - mae: 0.5921 - mape: 533212768.0000
Epoch 15/100
2/2 - 0s - loss: 0.3558 - mse: 0.3558 - mae: 0.5886 - mape: 528869952.0000
Epoch 16/100
2/2 - 0s - loss: 0.3510 - mse: 0.3510 - mae: 0.5852 - mape: 524590016.0000
Epoch 17/100
2/2 - 0s - loss: 0.3463 - mse: 0.3463 - mae: 0.5817 - mape: 520295264.0000
Epoch 18/100
2/2 - 0s - loss: 0.3417 - mse: 0.3417 - mae: 0.5783 - mape: 516026720.0000
Epoch 19/100
2/2 - 0s - loss: 0.3370 - mse: 0.3370 - mae: 0.5747 - mape: 511705632.0000
Epoch 20/100
2/2 - 0s - loss: 0.3325 - mse: 0.3325 - mae: 0.5713 - mape: 507470848.0000
Epoch 21/100
2/2 - 0s - loss: 0.3280 - mse: 0.3280 - mae: 0.5679 - mape: 503210432.0000
Epoch 22/100
2/2 - 0s - loss: 0.3236 - mse: 0.3236 - mae: 0.5645 - mape: 498992768.0000
Epoch 23/100
2/2 - 0s - loss: 0.3193 - mse: 0.3193 - mae: 0.5611 - mape: 494764064.0000
Epoch 24/100
2/2 - 0s - loss: 0.3150 - mse: 0.3150 - mae: 0.5577 - mape: 490546048.0000
Epoch 25/100
2/2 - 0s - loss: 0.3108 - mse: 0.3108 - mae: 0.5543 - mape: 486364960.0000
Epoch 26/100
```

```
2/2 - 0s - loss: 0.1672 - mse: 0.1672 - mae: 0.3915 - mape: 284575648.0000

Epoch 96/100

2/2 - 0s - loss: 0.1666 - mse: 0.1666 - mae: 0.3903 - mape: 283030912.0000

Epoch 97/100

2/2 - 0s - loss: 0.1659 - mse: 0.1659 - mae: 0.3890 - mape: 281487136.0000

Epoch 98/100

2/2 - 0s - loss: 0.1653 - mse: 0.1653 - mae: 0.3878 - mape: 279991040.0000

Epoch 99/100

2/2 - 0s - loss: 0.1647 - mse: 0.1647 - mae: 0.3866 - mape: 278499968.0000

Epoch 100/100

2/2 - 0s - loss: 0.1641 - mse: 0.1641 - mae: 0.3854 - mape: 277030912.0000
```

MSE 0.1641

Hyperparameter - Updating learning rate with different values

learning rate list = [0.001, 0.010, 0.1, 1]

```
print(f" Processing - learning rate changes { l_r }")
    node = 16 # No. of nodes - fixed
    model.add(Dense(node , input_shape=(31,), activation=fn))
    model.add(Dense(node -4, activation=fn))
    model.add(Dense(node - 8, activation=fn))
    model.add(Dense(1, activation=fn))
```

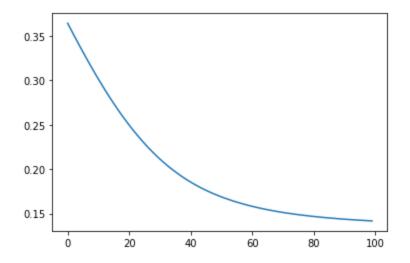
Processing - learning rate changes 0.001 Model: "sequential 75"

Layer (type)	Output Shape	Param #
dense_340 (Dense)	(None, 60)	1920
dense_341 (Dense)	(None, 56)	3416
dense_342 (Dense)	(None, 52)	2964

```
dense 343 (Dense)
                                                        53
                             (None, 1)
dense 344 (Dense)
                             (None, 16)
                                                        32
dense 345 (Dense)
                             (None, 12)
                                                        204
dense 346 (Dense)
                                                        104
                             (None, 8)
                                                        9
dense 347 (Dense)
                             (None, 1)
______
Epoch 1/100
2/2 - 0s - loss: 0.3643 - mse: 0.3643 - mae: 0.5947 - mape: 536381312.0000
Epoch 2/100
2/2 - 0s - loss: 0.3577 - mse: 0.3577 - mae: 0.5900 - mape: 530528768.0000
Epoch 3/100
2/2 - 0s - loss: 0.3510 - mse: 0.3510 - mae: 0.5852 - mape: 524621792.0000
Epoch 4/100
2/2 - 0s - loss: 0.3446 - mse: 0.3446 - mae: 0.5804 - mape: 518719520.0000
Epoch 5/100
2/2 - 0s - loss: 0.3381 - mse: 0.3381 - mae: 0.5756 - mape: 512749888.0000
Epoch 6/100
2/2 - 0s - loss: 0.3317 - mse: 0.3317 - mae: 0.5708 - mape: 506796032.0000
Epoch 7/100
2/2 - 0s - loss: 0.3255 - mse: 0.3255 - mae: 0.5660 - mape: 500818848.0000
Epoch 8/100
2/2 - 0s - loss: 0.3193 - mse: 0.3193 - mae: 0.5611 - mape: 494807712.0000
Epoch 9/100
2/2 - 0s - loss: 0.3132 - mse: 0.3132 - mae: 0.5563 - mape: 488821376.0000
Epoch 10/100
2/2 - 0s - loss: 0.3072 - mse: 0.3072 - mae: 0.5514 - ma
2/2 - 0s - loss: 0.1421 - mse: 0.1421 - mae: 0.3255 - mape: 202735664.0000
Epoch 99/100
2/2 - 0s - loss: 0.1419 - mse: 0.1419 - mae: 0.3246 - mape: 201703520.0000
Epoch 100/100
2/2 - 0s - loss: 0.1417 - mse: 0.1417 - mae: 0.3238 - mape: 200701552.0000
```

MSE = 0.3072

<tensorflow.python.keras.callbacks.History object at 0x7fde56103ac8>



Processing - learning rate changes 0.001

Model: "sequential_75"

Layer (typ	pe)	Output	Shape	Param #
dense_340	(Dense)	(None,	60)	1920
dense_341	(Dense)	(None,	56)	3416
dense_342	(Dense)	(None,	52)	2964
dense_343	(Dense)	(None,	1)	53
dense_344	(Dense)	(None,	16)	32
dense_345	(Dense)	(None,	12)	204
dense_346	(Dense)	(None,	8)	104
dense_347	(Dense)	(None,	1)	9
dense_348	(Dense)	(None,	16)	32
dense_349	(Dense)	(None,	12)	204
dense_350	(Dense)	(None,	8)	104

dense_351 (Dense) (None, 1) 9

Total params: 9,051 Trainable params: 9,051 Non-trainable params: 0

Epoch 1/100

2/2 - 0s - loss: 0.4698 - mse: 0.4698 - mae: 0.6625 - mape: 620697856.0000

Epoch 2/100

2/2 - 0s - loss: 0.4650 - mse: 0.4650 - mae: 0.6597 - mape: 617130432.0000

Epoch 3/100

2/2 - 0s - loss: 0.4599 - mse: 0.4599 - mae: 0.6566 - mape: 613437184.0000

Epoch 4/100

2/2 - 0s - loss: 0.4551 - mse: 0.4551 - mae: 0.6538 - mape: 609836480.0000

Epoch 5/100

2/2 - 0s - loss: 0.4501 - mse: 0.4501 - mae: 0.6508 - mape: 606132672.0000

Epoch 6/100

Processing - learning rate changes 0.01

Model: "sequential_75"

Layer (typ	pe)	Output	Shape	Param #
dense_340	(Dense)	(None,	60)	1920
dense_341	(Dense)	(None,	56)	3416
dense_342	(Dense)	(None,	52)	2964
dense_343	(Dense)	(None,	1)	53
dense_344	(Dense)	(None,	16)	32
dense_345	(Dense)	(None,	12)	204
dense_346	(Dense)	(None,	8)	104
dense_347	(Dense)	(None,	1)	9
dense_348	(Dense)	(None,	16)	32
dense_349	(Dense)	(None,	12)	204

dense_350 (Dense)

poch 1/100

2/2 - 0s - loss: 0.1378 - mse: 0.1378 - mae: 0.3135 - mape: 191614064.0000

```
Epoch 2/100
2/2 - 0s - loss: 0.1372 - mse: 0.1372 - mae: 0.3107 - mape: 188079936.0000
Epoch 3/100
2/2 - 0s - loss: 0.1367 - mse: 0.1367 - mae: 0.3078 - mape: 184612832.0000
Epoch 4/100
2/2 - 0s - loss: 0.1362 - mse: 0.1362 - mae: 0.3051 - mape: 181218944.0000
Epoch 5/100
2/2 - 0s - loss: 0.1358 - mse: 0.1358 - mae: 0.3024 - mape: 177920144.0000
Epoch 6/100
2/2 - 0s - loss: 0
2/2 - 0s - loss: 0.1329 - mse: 0.1329 - mae: 0.2649 - mape: 131811936.0000
Epoch 99/100
2/2 - 0s - loss: 0.1329 - mse: 0.1329 - mae: 0.2648 - mape: 131626288.0000
Epoch 100/100
2/2 - 0s - loss: 0.1329 - mse: 0.1329 - mae: 0.2646 - mape: 131436056.0000
<tensorflow.python.keras.callbacks.History object at 0x7fde5681cdd8>
```

dense_350	(Dense)	(None,	8)	104
dense_351	(Dense)	(None,	1)	9
dense_352	(Dense)	(None,	16)	32
dense_353	(Dense)	(None,	12)	204
dense_354	(Dense)	(None,	8)	104
dense_355	(Dense)	(None,	1)	9
dense_356	(Dense)	(None,	16)	32
dense_357	(Dense)	(None,	12)	204
dense_358	(Dense)	(None,	8)	104
dense_359	(Dense)	(None,	1)	9
dense_360	(Dense)	(None,	16)	32
dense_361	(Dense)	(None,	12)	204
dense_362	(Dense)	(None,	8)	104
dense_363	(Dense)	(None,	1)	9

Total params: 10,098 Trainable params: 10,098 Non-trainable params: 0

Epoch 1/100 2/2 - 0s - loss: 0.3325 - mse: 0.3325 - mae: 0.5717 - mape: 510276640.0000 Epoch 2/100 2/2 - 0s - loss: 0.3270 - mse: 0.3270 - mae: 0.5674 - mape: 505047744.0000 Epoch 3/100 2/2 - 0s - loss: 0.3216 - mse: 0.3216 - mae: 0.5632 - mape: 499807168.0000 Epoch 4/100 2/2 - 0s - loss: 0.3162 - mse: 0.3162 - mae: 0.5589 - mape: 494535232.0000 Epoch 5/100 2/2 - 0s - loss: 0.3108 - mse: 0.3108 - mae: 0.5546 - mape: 489254208.0000 Epoch 6/100 2/2 - 0s - loss: 0.3055 - mse: 0.3055 - mae: 0.5502 - mape: 483912384.0000 Epoch 7/100 2/2 - 0s - loss: 0.3003 - mse: 0.3003 - mae: 0.5459 - mape: 478637856.0000 Epoch 8/100 2/2 - 0s - loss: 0.2951 - mse: 0.2951 - mae: 0.5415 - mape: 473274368.0000 Epoch 9/100 2/2 - 0s - loss: 0.2901 - mse: 0.2901 - mae: 0.5372 - mape: 467965216.0000 Epoch 10/100 2/2 - 0s - loss: 0.2850 - mse: 0.2850 - mae: 0.5328 - mape: 462618080.0000 Epoch 11/100 2/2 - 0s - loss: 0.2802 - mse: 0.2802 - mae: 0.5285 - mape: 457305472.0000 Epoch 12/100 2/2 - 0s - loss: 0.2753 - mse: 0.2753 - mae: 0.5241 - mape: 451969216.0000 Epoch 13/100 2/2 - 0s - loss: 0.2706 - mse: 0.2706 - mae: 0.5198 - mape: 446660928.0000 Epoch 14/100 2/2 - 0s - loss: 0.2660 - mse: 0.2660 - mae: 0.5155 - mape: 441376192.0000 Epoch 15/100 2/2 - 0s - loss: 0.2614 - mse: 0.2614 - mae: 0.5111 - mape: 436100000.0000 Epoch 16/100 2/2 - 0s - loss: 0.2570 - mse: 0.2570 - mae: 0.5069 - mape: 430881888.0000 Epoch 17/100 2/2 - 0s - loss: 0.2526 - mse: 0.2526 - mae: 0.5026 - mape: 425654368.0000 Epoch 18/100 2/2 - 0s - loss: 0.2484 - mse: 0.2484 - mae: 0.4984 - mape: 420479680.0000 Epoch 19/100 2/2 - 0s - loss: 0.2442 - mse: 0.2442 - mae: 0.4941 - mape: 415317248.0000 Epoch 20/100 2/2 - 0s - loss: 0.2402 - mse: 0.2402 - mae: 0.4900 - mape: 410240704.0000 Epoch 21/100 2/2 - 0s - loss: 0.2363 - mse: 0.2363 - mae: 0.4859 - mape: 405179584.0000

```
Epoch 22/100
2/2 - 0s - loss: 0.2325 - mse: 0.2325 - mae: 0.4818 - mape: 400158720.0000
Epoch 23/100
2/2 - 0s - loss: 0.2288 - mse: 0.2288 - mae: 0.4777 - mape: 395178560.0000
Epoch 24/100
2/2 - 0s - loss: 0.2252 - mse: 0.2252 - mae: 0.4737 - mape: 390278432.0000
Epoch 25/100
2/2 - 0s - loss: 0.2217 - mse: 0.2217 - mae: 0.4698 - mape: 385448704.0000
Epoch 26/100
2/2 - 0s - loss: 0.2183 - mse: 0.2183 - mae: 0.4659 - mape: 380638080.0000
Epoch 27/100
2/2 - 0s - loss: 0.2150 - mse: 0.2150 - mae: 0.4620 - mape: 375906368.0000
Epoch 28/100
2/2 - 0s - loss: 0.2119 - mse: 0.2119 - mae: 0.4582 - mape: 371246848.0000
Epoch 29/100
2/2 - 0s - loss: 0.2088 - mse: 0.2088 - mae: 0.
MSE 0.2088
momentum list = [0.0, 0.2, 0.4, 0.9]
for momentum in momentum list:
     opt=keras.optimizers.SGD(lr=l r, momentum=momentum)
    print(f" Processing - momentum changes { momentum }")
     node = 16 # No. of nodes - fixed
    model.add(Dense(node , input shape=(31,), activation=fn))
    model.add(Dense(node -4, activation=fn))
     model.add(Dense(node - 8, activation=fn))
     model.add(Dense( 1, activation=fn))
poch 1/100
2/2 - 0s - loss: 0.3123 - mse: 0.3123 - mae: 0.5479 - mape: 483405568.0000
Epoch 2/100
2/2 - 0s - loss: 0.1458 - mse: 0.1458 - mae: 0.3518 - mape: 249633952.0000
Epoch 3/100
2/2 - 0s - loss: 0.1292 - mse: 0.1292 - mae: 0.3001 - mape: 187314096.0000
Epoch 4/100
2/2 - 0s - loss: 0.1258 - mse: 0.1258 - mae: 0.2809 - mape: 164267808.0000
```

Epoch 5/100 2/2 - 0s -

2/2 - 0s - loss: 0.1246 - mse: 0.1246 - mae: 0.2706 - mape: 151944464.0000

Epoch 6/100

2/2 - 0s - loss: 0.1240 - mse: 0.1240 - mae: 0.2643 - mape: 144419472.0000

Epoch 7/100

2/2 - 0s - loss: 0.1237 - mse: 0.1237 - mae: 0.2582 - mape: 136984800.0000

Epoch 8/100

2/2 - 0s - loss: 0.1236 - mse: 0.1236 - mae: 0.2569 - mape: 135468224.0000

Epoch 9/100

2/2 - 0s - loss: 0.1234 - mse: 0.1234 - mae: 0.2440 - mape: 120024056.0000

Epoch 100/100

2/2 - 0s - loss: 0.1234 - mse: 0.1234 - mae: 0.2448 - mape: 120934392.0000

poch 96/100

2/2 - 0s - loss: 0.1618 - mse: 0.1618 - mae: 0.1622 - mape: 313970.1875

Epoch 97/100

2/2 - 0s - loss: 0.1618 - mse: 0.1618 - mae: 0.1622 - mape: 317900.0312

Epoch 98/100

2/2 - 0s - loss: 0.1618 - mse: 0.1618 - mae: 0.1622 - mape: 321914.9062

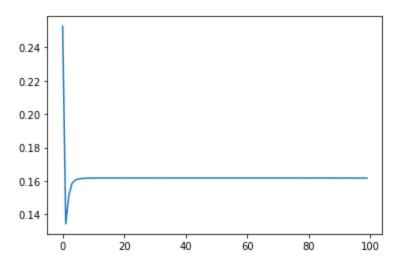
Epoch 99/100

2/2 - 0s - loss: 0.1618 - mse: 0.1618 - mae: 0.1622 - mape: 326058.9688

Epoch 100/100

2/2 - 0s - loss: 0.1618 - mse: 0.1618 - mae: 0.1622 - mape: 330319.1562 <tensorflow.python.keras.callbacks.History object at 0x7f039ccf16a0>

MSE 0.618



```
early stopping list = [ 'min', 'max', 'auto']
for e s in early stopping list:
print(f" Processing - early stopping changes { e_s }")
     node = 16 # No. of nodes - fixed
    model.add(Dense(node , input_shape=(31,), activation=fn))
    model.add(Dense(node -4, activation=fn))
    model.add(Dense(node - 8, activation=fn))
    model.add(Dense( 1, activation=fn))
    callback = k.callbacks.EarlyStopping(
    monitor='loss', min delta=0.0001,patience=1)
    1 r = 1
    momentum=0.9
    opt=keras.optimizers.SGD(lr=l r, momentum=momentum)
    history = model.fit(X_train, Y_train, validation_split=0.3,
epochs=100, callbacks=[callback])
    model.compile(optimizer=opt,loss='mse', metrics=['mse', 'mae',
'mape'])
odel: "sequential_17"
Layer (type)
              Output Shape
                            Param #
______
```

dense_64 (Dense)	(None, 60)	1920
dense_65 (Dense)	(None, 56)	3416
dense_66 (Dense)	(None, 52)	2964
dense_67 (Dense)	(None, 1)	53
dense_68 (Dense)	(None, 16)	32
dense_69 (Dense)	(None, 12)	204
dense_70 (Dense)	(None, 8)	104
dense_71 (Dense)	(None, 1)	9
dense_72 (Dense)	(None, 16)	32
dense_73 (Dense)	(None, 12)	204
dense_74 (Dense)	(None, 8)	104
dense_75 (Dense)	(None, 1)	9
dense_76 (Dense)	(None, 16)	32
dense_77 (Dense)	(None, 12)	204
dense_78 (Dense)	(None, 8)	104
dense_79 (Dense)	(None, 1)	9
dense_80 (Dense)	(None, 16)	32
dense_81 (Dense)	(None, 12)	204
dense_82 (Dense)	(None, 8)	104

dense_83 (Dense)	(None, 1)	9	
dense_84 (Dense)	(None, 16)	32	
dense_85 (Dense)	(None, 12)	204	
dense_86 (Dense)	(None, 8)	104	
dense_87 (Dense)	(None, 1)	9	
dense_88 (Dense)	(None, 16)	32	
dense_89 (Dense)	(None, 12)	204	
dense_90 (Dense)	(None, 8)	104	
dense_91 (Dense)	(None, 1)	9	
dense_92 (Dense)	(None, 16)	32	
dense_93 (Dense)	(None, 12)	204	
dense_94 (Dense)	(None, 8)	104	
dense_95 (Dense)	(None, 1)	9	
dense_96 (Dense)	(None, 16)	32	
dense_97 (Dense)	(None, 12)	204	
dense_98 (Dense)	(None, 8)	104	
dense_99 (Dense)	(None, 1)	9	
dense_100 (Dense)	(None, 16)	32	
dense_101 (Dense)	(None, 12)	204	

dense_102 (Dense)	(None, 8)	104	
dense_103 (Dense)	(None, 1)	9	
dense_104 (Dense)	(None, 16)	32	
dense_105 (Dense)	(None, 12)	204	
dense_106 (Dense)	(None, 8)	104	
dense_107 (Dense)	(None, 1)	9	
dense_108 (Dense)	(None, 16)	32	
dense_109 (Dense)	(None, 12)	204	
dense_110 (Dense)	(None, 8)	104	
dense_111 (Dense)	(None, 1)	9	
dense_112 (Dense)	(None, 16)	32	
dense_113 (Dense)	(None, 12)	204	
dense_114 (Dense)	(None, 8)	104	
dense_115 (Dense)	(None, 1)	9	
dense_116 (Dense)	(None, 16)	32	
dense_117 (Dense)	(None, 12)	204	
dense_118 (Dense)	(None, 8)	104	
dense_119 (Dense)	(None, 1)	9	
dense_120 (Dense)	(None, 16)	32	

dense_121 (Dense)	(None, 12)	204
dense_122 (Dense)	(None, 8)	104
dense_123 (Dense)	(None, 1)	9
dense_124 (Dense)	(None, 16)	32
dense_125 (Dense)	(None, 12)	204
dense_126 (Dense)	(None, 8)	104
dense_127 (Dense)	(None, 1)	9
dense_128 (Dense)	(None, 16)	32
dense_129 (Dense)	(None, 12)	204
dense_130 (Dense)	(None, 8)	104
dense_131 (Dense)	(None, 1)	9
dense_132 (Dense)	(None, 16)	32
dense_133 (Dense)	(None, 12)	204
dense_134 (Dense)	(None, 8)	104
dense_135 (Dense)	(None, 1)	9
dense_136 (Dense)	(None, 16)	32
dense_137 (Dense)	(None, 12)	204
dense_138 (Dense)	(None, 8)	104
dense_139 (Dense)	(None, 1)	9

dense_140 (Dense)	(None, 16)	32	
dense_141 (Dense)	(None, 12)	204	
dense_142 (Dense)	(None, 8)	104	
dense_143 (Dense)	(None, 1)	9	
dense_144 (Dense)	(None, 16)	32	
dense_145 (Dense)	(None, 12)	204	
dense_146 (Dense)	(None, 8)	104	
dense_147 (Dense)	(None, 1)	9	
dense_148 (Dense)	(None, 16)	32	
dense_149 (Dense)	(None, 12)	204	
dense_150 (Dense)	(None, 8)	104	
dense_151 (Dense)	(None, 1)	9	
dense_152 (Dense)	(None, 16)	32	
dense_153 (Dense)	(None, 12)	204	
dense_154 (Dense)	(None, 8)	104	
dense_155 (Dense)	(None, 1)	9	
dense_156 (Dense)	(None, 16)	32	
dense_157 (Dense)	(None, 12)	204	
dense_158 (Dense)	(None, 8)	104	

dense_159 (Dense)	(None, 1)	9	
dense_160 (Dense)	(None, 16)	32	
dense_161 (Dense)	(None, 12)	204	
dense_162 (Dense)	(None, 8)	104	
dense_163 (Dense)	(None, 1)	9	
dense_164 (Dense)	(None, 16)	32	
dense_165 (Dense)	(None, 12)	204	
dense_166 (Dense)	(None, 8)	104	
dense_167 (Dense)	(None, 1)	9	
dense_168 (Dense)	(None, 16)	32	
dense_169 (Dense)	(None, 12)	204	
dense_170 (Dense)	(None, 8)	104	
dense_171 (Dense)	(None, 1)	9	
dense_172 (Dense)	(None, 16)	32	
dense_173 (Dense)	(None, 12)	204	
dense_174 (Dense)	(None, 8)	104	
dense_175 (Dense)	(None, 1)	9	
dense_176 (Dense)	(None, 16)	32	
dense_177 (Dense)	(None, 12)	204	

dense_178 (Dense)	(None, 8)	104	
dense_179 (Dense)	(None, 1)	9	
dense_180 (Dense)	(None, 16)	32	
dense_181 (Dense)	(None, 12)	204	
dense_182 (Dense)	(None, 8)	104	
dense_183 (Dense)	(None, 1)	9	
dense_184 (Dense)	(None, 16)	32	
dense_185 (Dense)	(None, 12)	204	
dense_186 (Dense)	(None, 8)	104	
dense_187 (Dense)	(None, 1)	9	
dense_188 (Dense)	(None, 16)	32	
dense_189 (Dense)	(None, 12)	204	
dense_190 (Dense)	(None, 8)	104	
dense_191 (Dense)	(None, 1)	9	
=======================================			

Total params: 19,172 Trainable params: 19,172 Non-trainable params: 0

Epoch 1/100

17/17 [==============] - 1s 57ms/step - loss: 0.1875 - mse: 0.1875 - mae: 0.1960 - mape: 6989005.0000 - val_loss: 0.3391 - val_mse: 0.3391 - val_mae: 0.5768 - val_mape:

516953184.0000

Epoch 2/100

516783136.0000

Epoch 3/100

0.3089 - mape: 155124736.0000 - val_loss: 0.3390 - val_mse: 0.3390 - val_mae: 0.5767 -

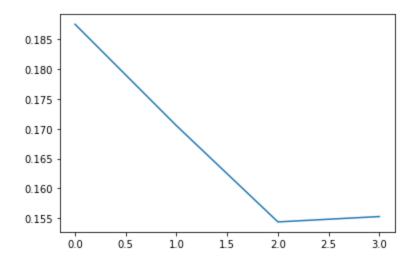
val_mape: 516838272.0000

Epoch 4/100

0.3190 - mape: 167551120.0000 - val_loss: 0.3390 - val_mse: 0.3390 - val_mae: 0.5767 -

val_mape: 516856224.0000

<tensorflow.python.keras.callbacks.History object at 0x7f037768d710>



Layer (type)	Output Shape	Param #	
dense_64 (Dense)	(None, 60)	1920	
dense_65 (Dense)	(None, 56)	3416	
dense_66 (Dense)	(None, 52)	2964	
dense_67 (Dense)	(None, 1)	53	
dense_68 (Dense)	(None, 16)	32	

dense_69 (Dense)	(None, 12)	204	
dense_70 (Dense)	(None, 8)	104	
dense_71 (Dense)	(None, 1)	9	
dense_72 (Dense)	(None, 16)	32	
dense_73 (Dense)	(None, 12)	204	
dense_74 (Dense)	(None, 8)	104	
dense_75 (Dense)	(None, 1)	9	
dense_76 (Dense)	(None, 16)	32	
dense_77 (Dense)	(None, 12)	204	
dense_78 (Dense)	(None, 8)	104	
dense_79 (Dense)	(None, 1)	9	
dense_80 (Dense)	(None, 16)	32	
dense_81 (Dense)	(None, 12)	204	
dense_82 (Dense)	(None, 8)	104	
dense_83 (Dense)	(None, 1)	9	
dense_84 (Dense)	(None, 16)	32	
dense_85 (Dense)	(None, 12)	204	
dense_86 (Dense)	(None, 8)	104	
dense_87 (Dense)	(None, 1)	9	

dense_88 (Dense)	(None, 16)	32	
dense_89 (Dense)	(None, 12)	204	
dense_90 (Dense)	(None, 8)	104	
dense_91 (Dense)	(None, 1)	9	
dense_92 (Dense)	(None, 16)	32	
dense_93 (Dense)	(None, 12)	204	
dense_94 (Dense)	(None, 8)	104	
dense_95 (Dense)	(None, 1)	9	
dense_96 (Dense)	(None, 16)	32	
dense_97 (Dense)	(None, 12)	204	
dense_98 (Dense)	(None, 8)	104	
dense_99 (Dense)	(None, 1)	9	
dense_100 (Dense)	(None, 16)	32	
dense_101 (Dense)	(None, 12)	204	
dense_102 (Dense)	(None, 8)	104	
dense_103 (Dense)	(None, 1)	9	
dense_104 (Dense)	(None, 16)	32	
dense_105 (Dense)	(None, 12)	204	
dense_106 (Dense)	(None, 8)	104	

dense_107 (Dense)	(None, 1)	9
dense_108 (Dense)	(None, 16)	32
dense_109 (Dense)	(None, 12)	204
dense_110 (Dense)	(None, 8)	104
dense_111 (Dense)	(None, 1)	9
dense_112 (Dense)	(None, 16)	32
dense_113 (Dense)	(None, 12)	204
dense_114 (Dense)	(None, 8)	104
dense_115 (Dense)	(None, 1)	9
dense_116 (Dense)	(None, 16)	32
dense_117 (Dense)	(None, 12)	204
dense_118 (Dense)	(None, 8)	104
dense_119 (Dense)	(None, 1)	9
dense_120 (Dense)	(None, 16)	32
dense_121 (Dense)	(None, 12)	204
dense_122 (Dense)	(None, 8)	104
dense_123 (Dense)	(None, 1)	9
dense_124 (Dense)	(None, 16)	32
dense_125 (Dense)	(None, 12)	204

dense_126 (Dense)	(None, 8)	104	
dense_127 (Dense)	(None, 1)	9	
dense_128 (Dense)	(None, 16)	32	
dense_129 (Dense)	(None, 12)	204	
dense_130 (Dense)	(None, 8)	104	
dense_131 (Dense)	(None, 1)	9	
dense_132 (Dense)	(None, 16)	32	
dense_133 (Dense)	(None, 12)	204	
dense_134 (Dense)	(None, 8)	104	
dense_135 (Dense)	(None, 1)	9	
dense_136 (Dense)	(None, 16)	32	
dense_137 (Dense)	(None, 12)	204	
dense_138 (Dense)	(None, 8)	104	
dense_139 (Dense)	(None, 1)	9	
dense_140 (Dense)	(None, 16)	32	
dense_141 (Dense)	(None, 12)	204	
dense_142 (Dense)	(None, 8)	104	
dense_143 (Dense)	(None, 1)	9	
dense_144 (Dense)	(None, 16)	32	

dense_145 (Dense)	(None, 12)	204	
dense_146 (Dense)	(None, 8)	104	
dense_147 (Dense)	(None, 1)	9	
dense_148 (Dense)	(None, 16)	32	
dense_149 (Dense)	(None, 12)	204	
dense_150 (Dense)	(None, 8)	104	
dense_151 (Dense)	(None, 1)	9	
dense_152 (Dense)	(None, 16)	32	
dense_153 (Dense)	(None, 12)	204	
dense_154 (Dense)	(None, 8)	104	
dense_155 (Dense)	(None, 1)	9	
dense_156 (Dense)	(None, 16)	32	
dense_157 (Dense)	(None, 12)	204	
dense_158 (Dense)	(None, 8)	104	
dense_159 (Dense)	(None, 1)	9	
dense_160 (Dense)	(None, 16)	32	
dense_161 (Dense)	(None, 12)	204	
dense_162 (Dense)	(None, 8)	104	
dense_163 (Dense)	(None, 1)	9	

dense_164 (Dense)	(None, 16)	32	
dense_165 (Dense)	(None, 12)	204	
dense_166 (Dense)	(None, 8)	104	
dense_167 (Dense)	(None, 1)	9	
dense_168 (Dense)	(None, 16)	32	
dense_169 (Dense)	(None, 12)	204	
dense_170 (Dense)	(None, 8)	104	
dense_171 (Dense)	(None, 1)	9	
dense_172 (Dense)	(None, 16)	32	
dense_173 (Dense)	(None, 12)	204	
dense_174 (Dense)	(None, 8)	104	
dense_175 (Dense)	(None, 1)	9	
dense_176 (Dense)	(None, 16)	32	
dense_177 (Dense)	(None, 12)	204	
dense_178 (Dense)	(None, 8)	104	
dense_179 (Dense)	(None, 1)	9	
dense_180 (Dense)	(None, 16)	32	
dense_181 (Dense)	(None, 12)	204	
dense_182 (Dense)	(None, 8)	104	

dense_183 (Dense)	(None, 1)	9
dense_184 (Dense)	(None, 16)	32
dense_185 (Dense)	(None, 12)	204
dense_186 (Dense)	(None, 8)	104
dense_187 (Dense)	(None, 1)	9
dense_188 (Dense)	(None, 16)	32
dense_189 (Dense)	(None, 12)	204
dense_190 (Dense)	(None, 8)	104
dense_191 (Dense)	(None, 1)	9
dense_192 (Dense)	(None, 16)	32
dense_193 (Dense)	(None, 12)	204
dense_194 (Dense)	(None, 8)	104
dense_195 (Dense)	(None, 1)	9
dense_196 (Dense)	(None, 16)	32
dense_197 (Dense)	(None, 12)	204
dense_198 (Dense)	(None, 8)	104
dense_199 (Dense)	(None, 1)	9
dense_200 (Dense)	(None, 16)	32
dense_201 (Dense)	(None, 12)	204

dense_202 (Dense)	(None, 8)	104
dense_203 (Dense)	(None, 1)	9

Total params: 20,219
Trainable params: 20,219
Non-trainable params: 0

Epoch 1/100

0.2789 - mape: 134038776.0000 - val_loss: 0.1338 - val_mse: 0.1338 - val_mae: 0.2813 -

val_mape: 151367200.0000

Epoch 2/100

0.2663 - mape: 119554168.0000 - val_loss: 0.1373 - val_mse: 0.1373 - val_mae: 0.3097 -

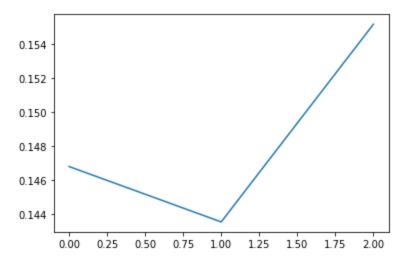
val_mape: 186410816.0000

Epoch 3/100

0.2649 - mape: 118791144.0000 - val_loss: 0.1438 - val_mse: 0.1438 - val_mae: 0.1966 -

val_mape: 47138352.0000

<tensorflow.python.keras.callbacks.History object at 0x7f03738116a0>



/usr/local/lib/python3.6/dist-packages/pandas/core/frame.py:4174: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation:

https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy

errors=errors,

Processing - early stopping changes auto

Model: "sequential_17"

	Output Shape		
dense_64 (Dense)		1920	
dense_65 (Dense)	(None, 56)	3416	
dense_66 (Dense)	(None, 52)	2964	
dense_67 (Dense)	(None, 1)	53	
dense_68 (Dense)	(None, 16)	32	
dense_69 (Dense)	(None, 12)	204	
dense_70 (Dense)	(None, 8)	104	
dense_71 (Dense)	(None, 1)	9	
dense_72 (Dense)	(None, 16)	32	
dense_73 (Dense)	(None, 12)	204	
dense_74 (Dense)	(None, 8)	104	
dense_75 (Dense)	(None, 1)	9	
dense_76 (Dense)	(None, 16)	32	
dense_77 (Dense)	(None, 12)	204	
dense_78 (Dense)	(None, 8)	104	

dense_79 (Dense)	(None, 1)	9	
dense_80 (Dense)	(None, 16)	32	
dense_81 (Dense)	(None, 12)	204	
dense_82 (Dense)	(None, 8)	104	
dense_83 (Dense)	(None, 1)	9	
dense_84 (Dense)	(None, 16)	32	
dense_85 (Dense)	(None, 12)	204	
dense_86 (Dense)	(None, 8)	104	
dense_87 (Dense)	(None, 1)	9	
dense_88 (Dense)	(None, 16)	32	
dense_89 (Dense)	(None, 12)	204	
dense_90 (Dense)	(None, 8)	104	
dense_91 (Dense)	(None, 1)	9	
dense_92 (Dense)	(None, 16)	32	
dense_93 (Dense)	(None, 12)	204	
dense_94 (Dense)	(None, 8)	104	
dense_95 (Dense)	(None, 1)	9	
dense_96 (Dense)	(None, 16)	32	
dense_97 (Dense)	(None, 12)	204	

dense_98 (Dense)	(None, 8)	104	
dense_99 (Dense)	(None, 1)	9	
dense_100 (Dense)	(None, 16)	32	
dense_101 (Dense)	(None, 12)	204	
dense_102 (Dense)	(None, 8)	104	
dense_103 (Dense)	(None, 1)	9	
dense_104 (Dense)	(None, 16)	32	
dense_105 (Dense)	(None, 12)	204	
dense_106 (Dense)	(None, 8)	104	
dense_107 (Dense)	(None, 1)	9	
dense_108 (Dense)	(None, 16)	32	
dense_109 (Dense)	(None, 12)	204	
dense_110 (Dense)	(None, 8)	104	
dense_111 (Dense)	(None, 1)	9	
dense_112 (Dense)	(None, 16)	32	
dense_113 (Dense)	(None, 12)	204	
dense_114 (Dense)	(None, 8)	104	
dense_115 (Dense)	(None, 1)	9	
dense_116 (Dense)	(None, 16)	32	

dense_117 (Dense)	(None, 12)	204	
dense_118 (Dense)	(None, 8)	104	
dense_119 (Dense)	(None, 1)	9	
dense_120 (Dense)	(None, 16)	32	
dense_121 (Dense)	(None, 12)	204	
dense_122 (Dense)	(None, 8)	104	
dense_123 (Dense)	(None, 1)	9	
dense_124 (Dense)	(None, 16)	32	
dense_125 (Dense)	(None, 12)	204	
dense_126 (Dense)	(None, 8)	104	
dense_127 (Dense)	(None, 1)	9	
dense_128 (Dense)	(None, 16)	32	
dense_129 (Dense)	(None, 12)	204	
dense_130 (Dense)	(None, 8)	104	
dense_131 (Dense)	(None, 1)	9	
dense_132 (Dense)	(None, 16)	32	
dense_133 (Dense)	(None, 12)	204	
dense_134 (Dense)	(None, 8)	104	
dense_135 (Dense)	(None, 1)	9	

dense_136 (Dense)	(None, 16)	32	
dense_137 (Dense)	(None, 12)	204	
dense_138 (Dense)	(None, 8)	104	
dense_139 (Dense)	(None, 1)	9	
dense_140 (Dense)	(None, 16)	32	
dense_141 (Dense)	(None, 12)	204	
dense_142 (Dense)	(None, 8)	104	
dense_143 (Dense)	(None, 1)	9	
dense_144 (Dense)	(None, 16)	32	
dense_145 (Dense)	(None, 12)	204	
dense_146 (Dense)	(None, 8)	104	
dense_147 (Dense)	(None, 1)	9	
dense_148 (Dense)	(None, 16)	32	
dense_149 (Dense)	(None, 12)	204	
dense_150 (Dense)	(None, 8)	104	
dense_151 (Dense)	(None, 1)	9	
dense_152 (Dense)	(None, 16)	32	
dense_153 (Dense)	(None, 12)	204	
dense_154 (Dense)	(None, 8)	104	

dense_155 (Dense)	(None, 1)	9	
dense_156 (Dense)	(None, 16)	32	
dense_157 (Dense)	(None, 12)	204	
dense_158 (Dense)	(None, 8)	104	
dense_159 (Dense)	(None, 1)	9	
dense_160 (Dense)	(None, 16)	32	
dense_161 (Dense)	(None, 12)	204	
dense_162 (Dense)	(None, 8)	104	
dense_163 (Dense)	(None, 1)	9	
dense_164 (Dense)	(None, 16)	32	
dense_165 (Dense)	(None, 12)	204	
dense_166 (Dense)	(None, 8)	104	
dense_167 (Dense)	(None, 1)	9	
dense_168 (Dense)	(None, 16)	32	
dense_169 (Dense)	(None, 12)	204	
dense_170 (Dense)	(None, 8)	104	
dense_171 (Dense)	(None, 1)	9	
dense_172 (Dense)	(None, 16)	32	
dense_173 (Dense)	(None, 12)	204	

dense_174 (Dense)	(None, 8)	104	
dense_175 (Dense)	(None, 1)	9	
dense_176 (Dense)	(None, 16)	32	
dense_177 (Dense)	(None, 12)	204	
dense_178 (Dense)	(None, 8)	104	
dense_179 (Dense)	(None, 1)	9	
dense_180 (Dense)	(None, 16)	32	
dense_181 (Dense)	(None, 12)	204	
dense_182 (Dense)	(None, 8)	104	
dense_183 (Dense)	(None, 1)	9	
dense_184 (Dense)	(None, 16)	32	
dense_185 (Dense)	(None, 12)	204	
dense_186 (Dense)	(None, 8)	104	
dense_187 (Dense)	(None, 1)	9	
dense_188 (Dense)	(None, 16)	32	
dense_189 (Dense)	(None, 12)	204	
dense_190 (Dense)	(None, 8)	104	
dense_191 (Dense)	(None, 1)	9	
dense_192 (Dense)	(None, 16)	32	

dense_193 (Dense)	(None, 12)	204	
dense_194 (Dense)	(None, 8)	104	
dense_195 (Dense)	(None, 1)	9	
dense_196 (Dense)	(None, 16)	32	
dense_197 (Dense)	(None, 12)	204	
dense_198 (Dense)	(None, 8)	104	
dense_199 (Dense)	(None, 1)	9	
dense_200 (Dense)	(None, 16)	32	
dense_201 (Dense)	(None, 12)	204	
dense_202 (Dense)	(None, 8)	104	
dense_203 (Dense)	(None, 1)	9	
dense_204 (Dense)	(None, 16)	32	
dense_205 (Dense)	(None, 12)	204	
dense_206 (Dense)	(None, 8)	104	
dense_207 (Dense)	(None, 1)	9	
		:=======	

Total params: 20,568
Trainable params: 20,568
Non-trainable params: 0

```
17/17 [=================] - 1s 62ms/step - loss: 0.1547 - mse: 0.1547 - mae: 0.3035 - mape: 151697712.0000 - val_loss: 0.1174 - val_mse: 0.1174 - val_mae: 0.2296 - val_mape: 111305184.0000

Epoch 2/100

17/17 [==================] - 0s 16ms/step - loss: 0.1610 - mse: 0.1610 - mae: 0.2928 - mape: 130924760.0000 - val_loss: 0.1186 - val_mse: 0.1186 - val_mae: 0.2611 - val_mape: 148672512.0000

<tensorflow.python.keras.callbacks.History object at 0x7f036ff9fa90>
```

Summary:

Trained the model by changing parameters:

Processing the model using Node = [12,24,36,48,60] obtained "Accuracy" [0.9995, 0.9995,0.9991, 1.000 and 0.9991] for MNIST Data set and "MSE" for Kaggle Data set[0.1578,0.1578,0.1483,0.1741,0.1381,0.1578]

Processing the model using

Tanh, Relu, and Logistic activation functions.

[Tanh, Relu and Logistic activation Fn] obtained 'Accuracy' [0.9981, 0.9991, 0.9981] and 'MSE" for Kaggle Data set [0.1394,0.1311,0.1338]

Processing the model using

```
Learning Rate [ 0.001, 0.010, 0.1, 1]
obtained "Accuracy' [ 0.9981, 0.9986, 0.9991,0.9986,]
and MSE" for Kaggle Data set [
0.136,0.1640,0.1421,0.1283,0.1316]
```

Processing the model using

Momentum [0.0, 0.2, 0.4, 0.9] obtained "Accuracy" [0.9995, 0.9986, 0.9981,0.9986] and "MSE" for Kaggle Data set [0.1234,0.1282,0.1456,0.1348]

Processing the model using

Early Stopping [min, max,auto] obtained "Accuracy" [0.9976, 0.9991,0.9986] and "MSE" for Kaggle Data set [0.1553,0.1673,0,1419]

Overall Accuracy is 100% with different inputs of the above parameters
And MSE is 0.1553.

Thank you