```
In [1]: import os
        import matplotlib.pyplot as plt
        import pandas as pd
        from yahoo fin import stock info as si
        class YahooStockData :
                def init (
                                self , company , from date , to date , save folder = "charts" ,
                                result folder = "results" , interval = '1d' ,
                                stock df = None
                                ) :
                        self. ticker = company
                        self. start date = from date
                        self.__end_date = to_date
                        self.interval = interval
                        self. result folder = f"results/{result folder}"
                        self.__stock_df = stock df
                        self. save chart folder = f"{self. result folder}/{save folder}"
                        self. create folder( self. save chart folder )
                def create folder( self , folder ) :
                        if not os.path.exists( folder ) :
                                os.makedirs (folder)
                def get stock data( self ) :
                        stock data = si.get data(
                                self. ticker , start date = self. start date ,
                                end date = self. end date ,
                                interval = self.interval
                                )
                        self. stock df = pd.DataFrame(      stock data.to records()).rename(
                                columns = { 'index' : 'date' }
                        self. stock df = self. stock df.set index( "date" )
                        self. stock df.sort index( ascending = True )
                        return self. stock df
                def plot stock data(
                                self , column = 'adjclose' , style = 'seaborn-dark' ,
                                figsize = (10 , 15)
                        if self. stock df is None :
                                self.get stock data()
                        plt.style.use( style )
                        self. stock df[ column ].plot( cmap = "viridis" , figsize = figsize )
                        plt.grid( )
                        file path = os.path.join(
                                self. save chart folder , f'{self. ticker} stock data .png'
                        plt.savefig( file path )
                        plt.show()
                        plt.close()
                def save stock data as pkl( self , file name , folder = "data sets" ) :
                        file path = f"{folder}/{file name}"
                        if self. stock df is None :
                                self.get stock data( )
                        if not os.path.exists( os.path.dirname( file path ) ) :
                                os.mkdir( os.path.dirname( file path ) )
                        self.__stock_df.to_pickle( file path )
                        return file path
```

```
self.__stock_df = pd.read_pickle( file_path )
                       return self. stock df
In [2]: import numpy as np
        class StockDataPreprocessor :
               def init (
                               self , stock history length = 50 , split ratio = 0.8 ,
                               normalise stock data = True
                       self. stock history length = stock history length
                       self. normalise stock data = normalise stock data
                       self. split ratio = split ratio
               def normalise data( self , data ) :
                       normalised data = [ ]
                       for window in data :
                               normalised_window = [ ((float( p ) / float( window[ 0 ] )) - 1)
                                                   p in window ]
                               normalised data.append( normalised window )
                       return normalised data
               def train test split( self , data to split ) :
                       split index = int( len( data to split ) * self. split ratio )
                       train data = data to split[ :split index ]
                       test data = data to split[ split index : ]
                       return train data , test data
               def extract y from data( self , data ) :
                       data = np.array( data )
                       inp = data[ : , :-1 ]
                       out = data[ : , -1 ]
                       inp = np.reshape( inp , (inp.shape[ 0 ] , inp.shape[ 1 ] , 1) )
                       return inp , out
               def prepare_data( self , data ) :
                       history length = self. stock history length + 1
                        nomalise data = [ ]
                       for index in range( len( data ) - history length ) :
                               nomalise data.append( data[ index : index + history length ] )
                       if self. normalise stock data :
                               nomalise data = self. normalise data( nomalise data)
                       x train , y train = self. extract y from data( train data)
                       x test , y test = self. extract y from data( test data )
                       return x train , y train , x test , y test
In [3]: from sklearn.ensemble import RandomForestRegressor
        from sklearn.metrics import mean absolute error , mean squared error
        from sklearn.model selection import train test split
        class RandomForestPredictor :
               def init ( self , n estimators = 100 , random state = 42 ) :
                       self. n estimators = n estimators
                       self.__random_state = random_state
                       self. model = RandomForestRegressor(
                              n estimators = self. n estimators ,
                               random state = self. random state
```

self. history = None

def load stock data from pkl(self , file path) :

```
x train , y train , test size = 0.2 , random state = 42
                        x train 2d = x train.reshape( x train.shape[ 0 ] , -1 )
                        x \text{ val } 2d = x \text{ val.reshape}(x \text{ val.shape}[0], -1)
                        history = {
                                         'loss' : [ ] , 'val loss' : [ ] , 'mae' : [ ] , 'mse' :
                        self. model.fit( x train 2d , y train )
                        __train_predictions = self.__model.predict( x_train_2d )
                         val predictions = self. model.predict( x val 2d )
                         train loss = mean squared error( y train , train predictions )
                          val loss = mean squared error( y val , val predictions )
                         train mae = mean absolute_error( y_train , __train_predictions )
                        history[ 'loss' ].append( train loss )
                        history[ 'val loss' ].append( val loss )
                        history[ 'mae' ].append( train mae )
                        history[ 'mse' ].append( val loss )
                        self. history = history
                        return history , self. model
In [4]:
        from keras.layers import LSTM
        from keras.layers.core import Activation , Dense , Dropout
        from keras.models import Sequential
        from keras.optimizers import RMSprop
        class LSTMPredictor :
                def init (
                                self , layers , learning rate = 0.0001 , loss = "mse" ,
                                metrics = [ 'mae' , 'mse' ] , batch_size = 512 , epochs = 100 ,
                                validation split = 0.05
                                ) :
                        self. layers = layers
                        self. metrics = metrics
                        self. learning rate = learning rate
                        self. loss = loss
                        self.__batch_size = batch_size
                        self. epochs = epochs
                        self. validation split = validation split
                        self. model = self. build model()
                def build model( self ) :
                        lstm model = Sequential()
                        length = len( self. layers )
                        for layer in self.__layers :
                                length -= 1
                                if length != 0 :
                                        lstm model.add(
                                                LSTM( units = layer , return sequences = True )
                                else :
                                        lstm model.add( LSTM( layer ) )
                                lstm model.add( Dropout( 0.2 ) )
                        lstm model.add( Dense( units = 1 ) )
                        lstm model.add( Activation( "linear" ) )
```

x train , x val , y train , y val = train test split(

def train(self , x train , y train) :

```
lstm model.compile(
                                loss = self. loss , optimizer = optimizer ,
                                metrics = self. metrics
                        return 1stm model
                def train( self , x train , y train ) :
                        history = self. model.fit(
                                        x train ,
                                        y train ,
                                        batch size = self. batch size ,
                                        epochs = self. epochs ,
                                        validation split = self. validation split
                        return history , self. model
In [5]: from keras.layers import GRU
        from keras.layers.core import Activation , Dense , Dropout
        from keras.models import Sequential
        from keras.optimizers import RMSprop
        class GRUPredictor :
                def init (
                                self , layers , learning rate = 0.0001 , loss = "mse" ,
                                metrics = [ 'mae' , 'mse' ] , batch size = 512 , epochs = 100 ,
                                validation split = 0.05
                                ) :
                        self. layers = layers
                        self. metrics = metrics
                        self. learning rate = learning rate
                        self.__loss = loss
                        self. batch size = batch size
                        self. epochs = epochs
                        self. validation split = validation split
                        self. model = self. build model()
                def build model( self ) :
                        gru model = Sequential()
                        length = len( self. layers )
                        for layer in self. layers :
                                length -= 1
                                if length != 0 :
                                        gru model.add( GRU( units = layer , return sequences = T
                                else :
                                        gru model.add( GRU( layer ) )
                                gru model.add( Dropout( 0.2 ) )
                        gru model.add( Dense( units = 1 ) )
                        gru model.add( Activation( "linear" ) )
                        optimizer = RMSprop( self. learning rate )
                        gru model.compile(
                                loss = self. loss , optimizer = optimizer ,
                                metrics = self. metrics
                        return gru model
                def train( self , x train , y train ) :
                        history = self. model.fit(
```

optimizer = RMSprop(self. learning rate)

```
from keras.layers import Conv1D , Flatten , MaxPooling1D
In [6]:
        from keras.layers.core import Dense
        from keras.models import Sequential
        from keras.optimizers import RMSprop
        class CNNPredictor :
                def init (
                                self , layers , input shape , learning rate = 0.0001 ,
                                filters = 64 , kernel size = 3 , pool size = 2 , epochs = 100 ,
                                loss = "mse" ,
                                metrics = [ 'mae' , 'mse' ]
                                ) :
                        self. layers = layers
                        self. filters = filters
                        self.__kernel_size = kernel_size
                        self. pool size = pool size
                        self. input shape = input shape
                        self. epochs = epochs
                        self.\_loss = loss
                        self. metrics = metrics
                        self. learning rate = learning rate
                        self. model = self. build model()
                def build model( self ) :
                        cnn model = Sequential()
                        cnn model.add(
                                Conv1D(
                                        filters = self. filters , kernel size = self. kernel s
                                        activation = 'relu' ,
                                        input shape = self. input shape
                        cnn model.add( MaxPooling1D( pool size = self. pool size ) )
                        cnn model.add( Flatten( ) )
                        for units in self. layers[1:]:
                                cnn model.add( Dense( units = units , activation = 'relu' ) )
                        optimizer = RMSprop( self. learning rate )
                        cnn model.compile(
                                optimizer = optimizer , loss = self. loss ,
                                metrics = self. metrics
                        return cnn model
                def train( self , x train , y train ) :
                        history = self.__model.fit(
                                        x train ,
                                        y train ,
                                        epochs = self. epochs ,
                                        validation split = 0.05
                        return history , self.__model
```

```
import matplotlib.pyplot as plt
import numpy as np
results folder = "results"
sub folders = [ f.path for f in os.scandir( results folder ) if f.is dir() ]
rmse values = {
                "Random Forest" : [ ] , "LSTM" : [ ] , "GRU" : [ ] , "CNN" : [ ]
epochs list = [ ]
seq lengths list = [ ]
def get metrics():
        for sub folder in sub folders :
                for file in os.listdir( sub folder ) :
                        if file.endswith(".json"):
                                results file = os.path.join( sub folder , file )
                                with open ( results file , "r" ) as f :
                                        results data = json.load(f)
                                for model in [ "Random Forest" , "LSTM" , "GRU" , "CNN"
                                        rmse values[ model ].append(
                                                        results data[ "all models" ][ "r
                                epochs = results data[ "train metadata" ][ "epochs" ]
                                seq len = results data[ "train metadata" ][ "seq len" ]
                                epochs list.append( epochs )
                                seq lengths list.append( seq len )
def plot and save metric and save():
       bar width = 0.15
        x = np.arange( len( epochs list ) )
       fig , ax = plt.subplots(figsize = (12, 6))
        ax.bar(
                        x - 3 * bar width / 2 , rmse values[ "Random Forest" ] ,
                        bar width ,
                        label = "Random Forest"
        ax.bar(
                        x - bar width / 2 , rmse values[ "LSTM" ] , bar width ,
                        label = "LSTM"
        ax.bar(
                       x + bar width / 2 , rmse values[ "GRU" ] , bar width , label =
                        "GRU"
        ax.bar(
                        x + 3 * bar width / 2 , rmse values[ "CNN" ] , bar width , label
                        "CNN"
        ax.set ylabel( "RMSE" )
        ax.set title(
                        "RMSE Comparison for Different Models, Epochs, and Sequence "
                        "Lengths"
        ax.set xticks( x )
        ax.set xticklabels(
                        [ f"E{e}, S{s}" for e , s in
                          zip( epochs list , seq lengths list ) ] , rotation = 45
        ax.legend()
        plt.savefig( "rmse comparison chart epochs seq lengths bar.png" )
        plt.show()
```

```
"Random Forest" : { "best" : float( "inf" ) } ,
                                "LSTM" : { "best" : float( "inf" ) , "worst" : 0 } ,
                                               : { "best" : float( "inf" ) , "worst" : 0 } ,
                                               : { "best" : float( "inf" ) , "worst" : 0 } ,
                                "CNN"
                for model in [ "Random Forest" , "LSTM" , "GRU" , "CNN" ] :
                        for i , rmse value in enumerate( rmse values[ model ] ) :
                                if model == "Random Forest" :
                                        best worst rmse[ model ][ "best" ] = rmse value
                                else :
                                        if rmse value < best worst rmse[ model ][ "best" ] :</pre>
                                                best worst rmse[ model ][ "best" ] = rmse value
                                                best worst rmse[ model ][ "best seq len" ] = \
                                                        seq lengths list[ i ]
                                                best worst rmse[ model ][ "best epochs" ] = epoc
                                        if rmse value > best worst rmse[ model ][ "worst" ] :
                                                best worst rmse[ model ][ "worst" ] = rmse value
                                                best worst rmse[ model ][ "worst seq len" ] = \
                                                        seq lengths list[ i ]
                                                best worst rmse[ model ][ "worst epochs" ] = epo
                                                        i 1
                final best model = min(
                                best worst rmse , key = lambda x : best worst rmse[ x ][ "best"
                best worst rmse[ "final best model" ] = final best model
                with open( "best worst rmse final good model.json" , "w" ) as f :
                        json.dump( best worst rmse , f , indent = 2 )
In [8]: import json
        import os
        import pickle
        import matplotlib.pyplot as plt
        import numpy as np
        from sklearn.metrics import mean absolute error , mean squared error
        class PerformanceUtil :
                def init (
                                self , histories , all predictions , y test , metadata ,
                                save folder = "charts" , result folder = "" ,
                                model folder = 'models'
                                ) :
                        self. histories = histories
                        self.__all_predictions = all predictions
                        self. y test = y test
                        self. result folder = f"results/{result folder}"
                        self. save chart folder = f"{self. result folder}/{save folder}"
                        self.__save_model_folder = f"{self.__result_folder}/{model_folder}"
                        self. metadata = metadata
                        self. error metrics = None
                        self. create folder( self. save model folder )
                        self. create folder( self. save chart folder )
                def create folder( self , folder ) :
                        if not os.path.exists( folder ) :
                                os.makedirs( folder )
```

def get final good model():

global best_worst_rmse
best worst rmse = {

```
def root absolute squared error(
                self , y true , y pred , threshold = 0.5
        return np.sqrt( mean absolute error( y true , y pred ) )
def root mean squared error( self , y true , y pred ) :
        return np.sqrt( mean squared error( y true , y pred ) )
def get error metrics( self , y test ) :
        performance metrics = [ 'rmse' , 'mae' ]
        self. error metrics = { metric : { } for metric in
                                 performance metrics }
        for model name , predictions in self. all predictions.items() :
                self. error metrics[ 'rmse' ][
                        model name ] = self. root mean squared error(
                        y test , predictions
                self. error metrics[ 'mae' ][
                        model name ] = self. root absolute squared error(
                        y test , predictions
def write result( self , best_model_name ) :
        data = {
                        "train metadata" : self. metadata ,
                        "all_models" : self.__error_metrics ,
"best_model" : best_model_name
        model file path = os.path.join(
                self. result folder , "model metrics with best models.json"
        with open( model file path , "w" ) as file :
                json.dump( data , file , indent = 4 )
def save models( self , models ) :
        for model name , model in models.items() :
                model file path = os.path.join(
                        self. save model folder , f'{model name} model.pkl'
                with open ( model file path , 'wb' ) as f :
                        pickle.dump( model , f )
def print and writebest model based on rmse( self ) :
        best model name = min(
                        self. error metrics[ 'rmse' ] ,
                        key = self. error metrics[ 'rmse' ].get
        best_rmse = self.__error_metrics[ 'rmse' ][ best_model_name ]
        print( "All Models and Root Mean Squared Error:" )
        for model name , rmse in self. error metrics[ 'rmse' ].items() :
                print( f"\t{model name}'s RMSE is {rmse:.4f}" )
        print(
                f"The best model is {best model name} with an RMSE of "
                f"{best rmse:.4f}"
        self. write result( best model name )
def plot loss curves( self ) :
        num models = len( self. histories )
        fig , axs = plt.subplots(
                num models , 1 , figsize = (10 , num models * 5) ,
                constrained layout = True
```

```
{f for} i , model history {f in} enumerate( self. histories ) :
                for model name , history in model history.items() :
                        axs[ i ].plot(
                                        history.history[ 'loss' ] , label = 'Tra
                        axs[ i ].plot(
                                        history.history[ 'val loss' ] ,
                                        label = 'Validation Loss'
                        axs[ i ].plot( history.history[ 'mse' ] , label = "MSE"
                        axs[ i ].plot( history.history[ 'mae' ] , label = "MAE"
                        axs[ i ].set title( f'Metrics of {model name}' )
                        axs[ i ].set xlabel( 'Epoch' )
                        axs[ i ].set ylabel( 'Loss' )
                        axs[ i ].legend( loc = 'upper right' )
        file path = os.path.join(
                self. save chart folder , 'all model loss curves.png'
        plt.savefig( file path )
        plt.show()
        plt.close()
def plot all models performance( self ) :
        for model history in self. histories :
                for model name , history in model history.items() :
                        plt.plot(
                                        history.history[ 'val loss' ] ,
                                        label = f"{model name} Validation"
        plt.title( 'All Models Validation Loss' )
        plt.xlabel( 'Epoch' )
       plt.ylabel( 'Loss' )
        plt.legend( loc = 'upper right' )
        file path = os.path.join(
                self. save chart folder , 'all model validation loss curves.png
        plt.savefig( file path )
        plt.show()
        plt.close()
def plot all predictions( self ) :
        plt.plot( self. y test , label = 'True Values' )
        for model name , predictions in self. all predictions.items():
                plt.plot( predictions , label = f'{model name} Predictions' )
        plt.title( 'All Models Predictions vs True Values' )
        plt.legend( loc = 'upper right' )
        plt.show()
        plt.close()
        n models = len( self. all predictions )
        fig , axes = plt.subplots(
                n models , 1 , figsize = (10 , n models * 5) , sharex = True
        for i , (model name , predictions) in enumerate(
                       self. all predictions.items()
                axes[ i ].plot(
                        self. y test , label = 'True Values' , alpha = 0.6
                axes[ i ].plot(
                        predictions , label = f'{model name} Predictions' , alph
                axes[ i ].set title( f'{model name} Predictions vs True Values'
                axes[ i ].legend( loc = 'upper right' )
```

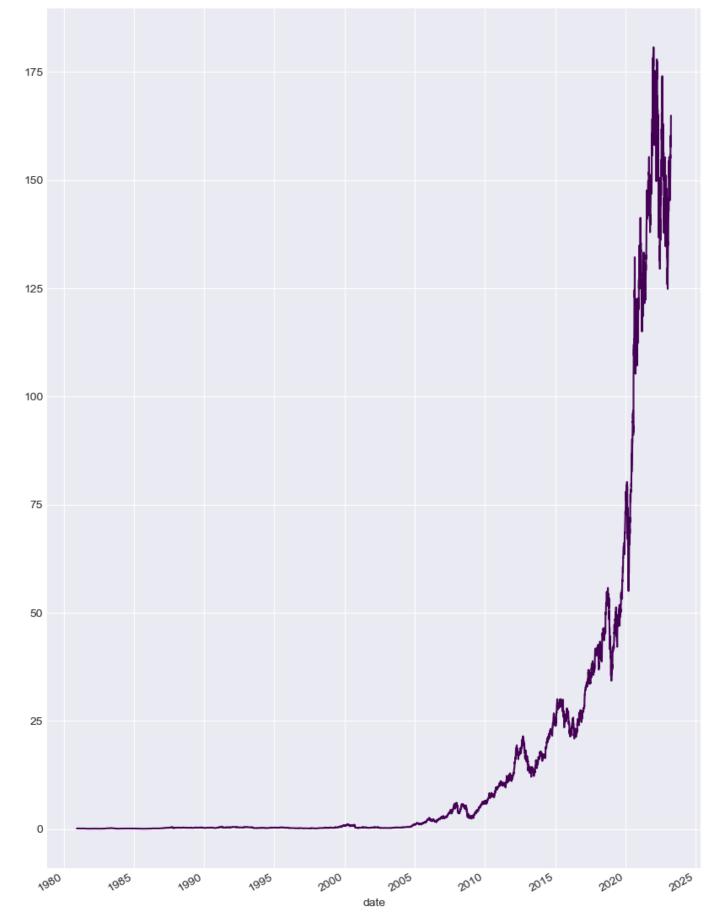
```
plt.xlabel( 'Time Steps' )
       plt.tight layout()
       file path = os.path.join(
               self. save chart folder ,
                'all model predictions vs true values.png'
       plt.savefig( file path )
       plt.show()
       plt.close( )
def plot error metric comparison( self ) :
       self. get error metrics( self. y test )
       fig , axs = plt.subplots(1, 2, figsize = (15, 5))
        for idx , (metric , model results) in enumerate(
                       self. error metrics.items()
               model names = list( model results.keys() )
               metric values = list( model results.values())
               axs[ idx ].bar( model names , metric values )
               axs[ idx ].set title( f'{metric.upper()} Comparison')
               axs[ idx ].set xlabel( 'Models')
               axs[ idx ].set ylabel( metric.upper( ) )
       file_path = os.path.join(
               self. save chart folder , 'all model error metric comparison.pn
       plt.savefig( file path )
       plt.show()
       plt.close()
       x train , y train , x test , y test , layers , epochs
```

```
In [10]: def train and evaluate models(
                         ) :
                 model histories = [ ]
                 rf predictor = RandomForestPredictor()
                 rf history , rf model = rf predictor.train( x train , y train )
                 lstm predictor = LSTMPredictor( layers , epochs = epochs )
                 lstm history , lstm model = lstm predictor.train( x train , y train )
                 model_histories.append( { "LSTM" : lstm history } )
                 gru predictor = GRUPredictor( layers , epochs = epochs )
                 gru history , gru model = gru predictor.train( x train , y train )
                 model histories.append( { "GRU" : gru history } )
                 cnn predictor = CNNPredictor(
                         layers , input shape = (x train.shape[ 1 ] , 1) , epochs = epochs
                 cnn history , cnn model = cnn predictor.train( x train , y train )
                 model histories.append( { "CNN" : cnn history } )
                 all predictions = {
                                  "Random Forest" : rf model.predict(
                                         x test.reshape( x test.shape[ 0 ] , -1 )
                                  "LSTM"
                                                 : lstm model.predict( x test ) ,
                                  "GRU"
                                                 : gru model.predict( x test ) ,
                                  "CNN"
                                                 : cnn model.predict( x test )
                 all models = {
```

```
"Random Forest" : rf model ,
                        "LSTM" : lstm_model ,
                        "GRU"
                                      : gru model ,
                        "CNN"
                                   : cnn model
       return model histories , all predictions , all models
def get and save yahoo stock data (
               ticker , start date , end date , result folder , file name
       yahoo stock data = YahooStockData(
               ticker , start date , end date , result folder = result folder
       stock df = yahoo stock data.get stock data()
       yahoo stock data.plot stock data()
       file path = yahoo stock data.save stock data as pkl( file name )
       stock df loaded = yahoo stock data.load stock data from pkl( file path )
       return stock df loaded
def main():
       start date = '1875-02-01 00:00:00'
       end date = '2023-04-01 23:59:59'
       ticker = "AAPL"
       file name = "apple.pkl"
       seq len = 15
       epochs = 200
       layers = [ seq len , 50 , 100 , 1 ]
       current time = datetime.datetime.now()
       time stamp = current time.strftime( "%d %H-%M-%S" )
       result folder = f"Ticker {ticker} {time stamp} layers" + " ".join(
                        [ str(i) for i in layers ]
                       ) + f" epochs {epochs}"
       stock df loaded = get and save yahoo stock data(
               ticker , start date , end date , result folder , file name
       data preprocessing = StockDataPreprocessor( seq len )
       stock data = stock df loaded[ 'adjclose' ].values
       x train , y train , x test , y test = data preprocessing.prepare data(
               stock data
       model histories , all predictions , all models = train and evaluate models(
               x train , y train , x test , y test , layers ,
       metadata = {
                        "trained at" : time stamp ,
                        "seq len" : seq len ,
                       "epochs" : epochs ,
"layers" : layers ,
                        "data"
                                    : {
                                        "ticker" : ticker , "start date" : start date ,
                                        "end date" : end date ,
                                        "count" : {
                                                        "x train" : len( x train ) ,
                                                        "y_train" : len( y_train ) ,
                                                        "x test" : len(x_test),
                                                        "y test" : len( y test)
```

/var/folders/ms/2f084nvx1n5b5wq69jt8j3cc0000gn/T/ipykernel_86416/49938427.py:45: Matplot libDeprecationWarning: The seaborn styles shipped by Matplotlib are deprecated since 3. 6, as they no longer correspond to the styles shipped by seaborn. However, they will rem ain available as 'seaborn-v0_8-<style>'. Alternatively, directly use the seaborn API ins tead.

plt.style.use(style)



Metal device set to: Apple M1 Max Epoch 1/200

2023-05-09 18:19:51.644425: I tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:306] Could not identify NUMA node of platform GPU ID 0, defaulting to 0. Your kernel may not have been built with NUMA support.

2023-05-09 18:19:51.644543: I tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:272] Created TensorFlow device (/job:localhost/replica:0/task:0/devic e:GPU:0 with 0 MB memory) -> physical PluggableDevice (device: 0, name: METAL, pci bus i

```
d: <undefined>)
2023-05-09 18:19:51.740737: W tensorflow/core/platform/profile utils/cpu utils.cc:128] F
ailed to get CPU frequency: 0 Hz
2023-05-09 18:19:53.717652: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.015720: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.134512: I tensorflow/core/grappler/optimizers/custom_graph_optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.230866: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.318433: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.515596: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.662493: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.833315: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:54.970530: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
134
2023-05-09 18:19:56.568293: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:56.667115: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:56.727730: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
e: 0.0134 - val loss: 0.0029 - val mae: 0.0417 - val mse: 0.0029
Epoch 2/200
2023-05-09 18:19:56.794079: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:19:56.884656: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
e: 0.0121 - val loss: 0.0026 - val mae: 0.0397 - val mse: 0.0026
Epoch 3/200
e: 0.0109 - val loss: 0.0023 - val mae: 0.0374 - val mse: 0.0023
Epoch 4/200
e: 0.0097 - val loss: 0.0022 - val mae: 0.0355 - val mse: 0.0022
e: 0.0090 - val loss: 0.0020 - val mae: 0.0351 - val mse: 0.0020
Epoch 6/200
e: 0.0087 - val loss: 0.0020 - val mae: 0.0348 - val mse: 0.0020
Epoch 7/200
e: 0.0084 - val loss: 0.0019 - val mae: 0.0345 - val mse: 0.0019
e: 0.0082 - val loss: 0.0019 - val mae: 0.0342 - val mse: 0.0019
Epoch 9/200
e: 0.0082 - val loss: 0.0018 - val mae: 0.0338 - val mse: 0.0018
Epoch 10/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0080 - mae: 0.0661 - ms
e: 0.0080 - val loss: 0.0018 - val mae: 0.0335 - val mse: 0.0018
Epoch 11/200
e: 0.0078 - val loss: 0.0018 - val mae: 0.0331 - val mse: 0.0018
```

```
Epoch 12/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0077 - mae: 0.0649 - ms
e: 0.0077 - val loss: 0.0018 - val mae: 0.0328 - val mse: 0.0018
Epoch 13/200
e: 0.0075 - val loss: 0.0018 - val mae: 0.0327 - val mse: 0.0018
Epoch 14/200
e: 0.0075 - val loss: 0.0016 - val mae: 0.0317 - val mse: 0.0016
Epoch 15/200
e: 0.0072 - val loss: 0.0016 - val mae: 0.0312 - val mse: 0.0016
Epoch 16/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0070 - mae: 0.0615 - ms
e: 0.0070 - val loss: 0.0016 - val mae: 0.0309 - val mse: 0.0016
Epoch 17/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0068 - mae: 0.0604 - ms
e: 0.0068 - val loss: 0.0015 - val mae: 0.0302 - val mse: 0.0015
Epoch 18/200
e: 0.0066 - val loss: 0.0014 - val mae: 0.0297 - val mse: 0.0014
Epoch 19/200
e: 0.0066 - val loss: 0.0014 - val mae: 0.0289 - val mse: 0.0014
Epoch 20/200
e: 0.0067 - val loss: 0.0014 - val mae: 0.0284 - val mse: 0.0014
Epoch 21/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0064 - mae: 0.0578 - ms
e: 0.0064 - val loss: 0.0013 - val mae: 0.0278 - val mse: 0.0013
Epoch 22/200
e: 0.0064 - val loss: 0.0013 - val mae: 0.0273 - val mse: 0.0013
Epoch 23/200
e: 0.0062 - val loss: 0.0012 - val mae: 0.0268 - val mse: 0.0012
Epoch 24/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0061 - mae: 0.0563 - ms
e: 0.0061 - val loss: 0.0012 - val mae: 0.0265 - val mse: 0.0012
Epoch 25/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0060 - mae: 0.0556 - ms
e: 0.0060 - val loss: 0.0012 - val mae: 0.0263 - val mse: 0.0012
Epoch 26/200
e: 0.0059 - val loss: 0.0012 - val mae: 0.0260 - val mse: 0.0012
Epoch 27/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0058 - mae: 0.0549 - ms
e: 0.0058 - val loss: 0.0012 - val mae: 0.0258 - val mse: 0.0012
Epoch 28/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0058 - mae: 0.0548 - ms
e: 0.0058 - val loss: 0.0012 - val mae: 0.0259 - val mse: 0.0012
Epoch 29/200
e: 0.0059 - val loss: 0.0011 - val mae: 0.0254 - val mse: 0.0011
Epoch 30/200
e: 0.0057 - val loss: 0.0011 - val mae: 0.0258 - val mse: 0.0011
Epoch 31/200
e: 0.0059 - val loss: 0.0011 - val mae: 0.0250 - val mse: 0.0011
Epoch 32/200
e: 0.0059 - val loss: 0.0011 - val mae: 0.0246 - val mse: 0.0011
Epoch 33/200
```

e: 0.0056 - val loss: 0.0011 - val mae: 0.0245 - val mse: 0.0011

```
Epoch 34/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0056 - mae: 0.0534 - ms
e: 0.0056 - val loss: 0.0011 - val mae: 0.0247 - val mse: 0.0011
Epoch 35/200
e: 0.0056 - val loss: 0.0010 - val mae: 0.0240 - val mse: 0.0010
Epoch 36/200
e: 0.0056 - val loss: 9.9940e-04 - val mae: 0.0238 - val mse: 9.9940e-04
Epoch 37/200
e: 0.0055 - val loss: 9.9410e-04 - val mae: 0.0237 - val mse: 9.9410e-04
Epoch 38/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0054 - mae: 0.0522 - ms
e: 0.0054 - val loss: 9.9141e-04 - val mae: 0.0237 - val mse: 9.9141e-04
Epoch 39/200
16/16 [=============== ] - 1s 44ms/step - loss: 0.0056 - mae: 0.0523 - ms
e: 0.0056 - val loss: 9.6017e-04 - val mae: 0.0232 - val mse: 9.6017e-04
Epoch 40/200
e: 0.0054 - val loss: 9.3382e-04 - val mae: 0.0229 - val mse: 9.3382e-04
Epoch 41/200
e: 0.0052 - val loss: 9.1400e-04 - val mae: 0.0226 - val mse: 9.1400e-04
Epoch 42/200
e: 0.0053 - val loss: 9.0442e-04 - val mae: 0.0224 - val mse: 9.0442e-04
Epoch 43/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0053 - mae: 0.0510 - ms
e: 0.0053 - val loss: 8.8410e-04 - val mae: 0.0223 - val mse: 8.8410e-04
Epoch 44/200
e: 0.0052 - val loss: 8.7978e-04 - val mae: 0.0221 - val mse: 8.7978e-04
Epoch 45/200
e: 0.0052 - val loss: 8.8932e-04 - val mae: 0.0224 - val mse: 8.8932e-04
Epoch 46/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0050 - mae: 0.0498 - ms
e: 0.0050 - val loss: 8.4935e-04 - val mae: 0.0217 - val mse: 8.4935e-04
Epoch 47/200
e: 0.0053 - val loss: 8.2921e-04 - val mae: 0.0215 - val mse: 8.2921e-04
Epoch 48/200
16/16 [=================== ] - 1s 40ms/step - loss: 0.0050 - mae: 0.0496 - ms
e: 0.0050 - val loss: 8.0150e-04 - val mae: 0.0211 - val mse: 8.0150e-04
Epoch 49/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0049 - mae: 0.0492 - ms
e: 0.0049 - val loss: 7.9289e-04 - val mae: 0.0210 - val mse: 7.9289e-04
Epoch 50/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0050 - mae: 0.0492 - ms
e: 0.0050 - val loss: 7.7968e-04 - val mae: 0.0208 - val mse: 7.7968e-04
Epoch 51/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0050 - mae: 0.0486 - ms
e: 0.0050 - val loss: 7.7005e-04 - val mae: 0.0207 - val mse: 7.7005e-04
Epoch 52/200
e: 0.0049 - val loss: 7.6227e-04 - val mae: 0.0205 - val mse: 7.6227e-04
Epoch 53/200
e: 0.0050 - val loss: 7.3821e-04 - val mae: 0.0203 - val mse: 7.3821e-04
Epoch 54/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0049 - mae: 0.0484 - ms
e: 0.0049 - val_loss: 7.2992e-04 - val_mae: 0.0202 - val_mse: 7.2992e-04
Epoch 55/200
```

e: 0.0050 - val loss: 7.2690e-04 - val mae: 0.0201 - val mse: 7.2690e-04

```
Epoch 56/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0048 - mae: 0.0480 - ms
e: 0.0048 - val loss: 7.0269e-04 - val mae: 0.0199 - val mse: 7.0269e-04
Epoch 57/200
e: 0.0049 - val loss: 7.0419e-04 - val mae: 0.0198 - val mse: 7.0419e-04
Epoch 58/200
e: 0.0047 - val loss: 6.8384e-04 - val mae: 0.0196 - val mse: 6.8384e-04
Epoch 59/200
e: 0.0048 - val loss: 6.8906e-04 - val mae: 0.0196 - val mse: 6.8906e-04
Epoch 60/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0048 - mae: 0.0471 - ms
e: 0.0048 - val loss: 6.8280e-04 - val mae: 0.0197 - val mse: 6.8280e-04
Epoch 61/200
e: 0.0044 - val loss: 6.6882e-04 - val mae: 0.0194 - val mse: 6.6882e-04
Epoch 62/200
e: 0.0051 - val loss: 6.6675e-04 - val mae: 0.0194 - val mse: 6.6675e-04
Epoch 63/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0046 - mae: 0.0465 - ms
e: 0.0046 - val loss: 6.6514e-04 - val mae: 0.0195 - val mse: 6.6514e-04
Epoch 64/200
e: 0.0047 - val loss: 6.7575e-04 - val mae: 0.0193 - val mse: 6.7575e-04
Epoch 65/200
16/16 [=================== ] - 1s 40ms/step - loss: 0.0048 - mae: 0.0469 - ms
e: 0.0048 - val loss: 6.4104e-04 - val mae: 0.0191 - val mse: 6.4104e-04
Epoch 66/200
e: 0.0046 - val loss: 6.3209e-04 - val mae: 0.0189 - val mse: 6.3209e-04
Epoch 67/200
e: 0.0045 - val loss: 6.3556e-04 - val mae: 0.0188 - val mse: 6.3556e-04
Epoch 68/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0047 - mae: 0.0464 - ms
e: 0.0047 - val loss: 6.2673e-04 - val mae: 0.0187 - val mse: 6.2673e-04
Epoch 69/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0045 - mae: 0.0458 - ms
e: 0.0045 - val loss: 6.1586e-04 - val mae: 0.0187 - val mse: 6.1586e-04
Epoch 70/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0046 - mae: 0.0459 - ms
e: 0.0046 - val loss: 6.3506e-04 - val mae: 0.0187 - val mse: 6.3506e-04
Epoch 71/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0045 - mae: 0.0457 - ms
e: 0.0045 - val loss: 6.1962e-04 - val mae: 0.0188 - val mse: 6.1962e-04
Epoch 72/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0046 - mae: 0.0456 - ms
e: 0.0046 - val loss: 5.9767e-04 - val mae: 0.0184 - val mse: 5.9767e-04
Epoch 73/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0045 - mae: 0.0453 - ms
e: 0.0045 - val loss: 5.9305e-04 - val mae: 0.0183 - val mse: 5.9305e-04
Epoch 74/200
e: 0.0044 - val loss: 5.9717e-04 - val mae: 0.0184 - val mse: 5.9717e-04
Epoch 75/200
e: 0.0045 - val loss: 5.8988e-04 - val_mae: 0.0183 - val_mse: 5.8988e-04
Epoch 76/200
e: 0.0044 - val_loss: 5.8297e-04 - val_mae: 0.0180 - val_mse: 5.8297e-04
Epoch 77/200
```

e: 0.0044 - val loss: 5.8346e-04 - val mae: 0.0180 - val mse: 5.8346e-04

```
Epoch 78/200
e: 0.0046 - val loss: 5.7137e-04 - val mae: 0.0179 - val mse: 5.7137e-04
Epoch 79/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0044 - mae: 0.0446 - ms
e: 0.0044 - val loss: 5.7516e-04 - val mae: 0.0178 - val mse: 5.7516e-04
Epoch 80/200
e: 0.0045 - val loss: 5.6992e-04 - val mae: 0.0178 - val mse: 5.6992e-04
Epoch 81/200
e: 0.0045 - val loss: 5.5249e-04 - val mae: 0.0175 - val mse: 5.5249e-04
Epoch 82/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0045 - mae: 0.0445 - ms
e: 0.0045 - val loss: 5.5981e-04 - val mae: 0.0176 - val mse: 5.5981e-04
Epoch 83/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0045 - mae: 0.0444 - ms
e: 0.0045 - val loss: 5.6937e-04 - val mae: 0.0179 - val mse: 5.6937e-04
Epoch 84/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0043 - mae: 0.0440 - ms
e: 0.0043 - val loss: 5.4663e-04 - val mae: 0.0174 - val mse: 5.4663e-04
Epoch 85/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0048 - mae: 0.0449 - ms
e: 0.0048 - val loss: 5.5983e-04 - val mae: 0.0177 - val mse: 5.5983e-04
Epoch 86/200
e: 0.0044 - val loss: 5.4608e-04 - val mae: 0.0173 - val mse: 5.4608e-04
Epoch 87/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0044 - mae: 0.0441 - ms
e: 0.0044 - val loss: 5.4746e-04 - val mae: 0.0174 - val mse: 5.4746e-04
Epoch 88/200
e: 0.0044 - val loss: 5.3776e-04 - val mae: 0.0174 - val mse: 5.3776e-04
Epoch 89/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0043 - mae: 0.0440 - ms
e: 0.0043 - val loss: 5.3567e-04 - val mae: 0.0172 - val mse: 5.3567e-04
Epoch 90/200
e: 0.0043 - val loss: 5.2640e-04 - val mae: 0.0172 - val mse: 5.2640e-04
Epoch 91/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0043 - mae: 0.0436 - ms
e: 0.0043 - val loss: 5.2310e-04 - val mae: 0.0170 - val mse: 5.2310e-04
Epoch 92/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0043 - mae: 0.0437 - ms
e: 0.0043 - val loss: 5.2252e-04 - val mae: 0.0169 - val mse: 5.2252e-04
Epoch 93/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0043 - mae: 0.0440 - ms
e: 0.0043 - val loss: 5.2780e-04 - val mae: 0.0171 - val mse: 5.2780e-04
Epoch 94/200
e: 0.0042 - val loss: 5.4014e-04 - val mae: 0.0171 - val mse: 5.4014e-04
Epoch 95/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0044 - mae: 0.0435 - ms
e: 0.0044 - val loss: 5.2448e-04 - val mae: 0.0170 - val mse: 5.2448e-04
Epoch 96/200
e: 0.0043 - val loss: 5.1874e-04 - val mae: 0.0168 - val mse: 5.1874e-04
Epoch 97/200
e: 0.0041 - val loss: 5.0376e-04 - val mae: 0.0166 - val mse: 5.0376e-04
Epoch 98/200
e: 0.0043 - val loss: 5.1234e-04 - val mae: 0.0167 - val mse: 5.1234e-04
Epoch 99/200
```

e: 0.0044 - val loss: 5.0290e-04 - val mae: 0.0165 - val mse: 5.0290e-04

```
Epoch 100/200
e: 0.0043 - val loss: 5.0134e-04 - val mae: 0.0165 - val mse: 5.0134e-04
Epoch 101/200
e: 0.0041 - val loss: 4.8738e-04 - val mae: 0.0163 - val mse: 4.8738e-04
Epoch 102/200
e: 0.0042 - val loss: 4.9741e-04 - val mae: 0.0165 - val mse: 4.9741e-04
Epoch 103/200
e: 0.0045 - val loss: 5.0834e-04 - val mae: 0.0166 - val mse: 5.0834e-04
Epoch 104/200
16/16 [============== ] - 1s 38ms/step - loss: 0.0040 - mae: 0.0424 - ms
e: 0.0040 - val loss: 4.8503e-04 - val mae: 0.0164 - val mse: 4.8503e-04
Epoch 105/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0043 - mae: 0.0425 - ms
e: 0.0043 - val loss: 4.9818e-04 - val mae: 0.0165 - val mse: 4.9818e-04
Epoch 106/200
e: 0.0040 - val loss: 4.8166e-04 - val mae: 0.0162 - val mse: 4.8166e-04
Epoch 107/200
e: 0.0043 - val loss: 4.9139e-04 - val mae: 0.0165 - val mse: 4.9139e-04
Epoch 108/200
e: 0.0042 - val loss: 4.7525e-04 - val mae: 0.0161 - val mse: 4.7525e-04
Epoch 109/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0043 - mae: 0.0428 - ms
e: 0.0043 - val loss: 4.7916e-04 - val mae: 0.0162 - val mse: 4.7916e-04
Epoch 110/200
e: 0.0041 - val loss: 4.6966e-04 - val mae: 0.0161 - val mse: 4.6966e-04
Epoch 111/200
e: 0.0041 - val loss: 4.5843e-04 - val mae: 0.0158 - val mse: 4.5843e-04
Epoch 112/200
e: 0.0040 - val loss: 4.8154e-04 - val mae: 0.0161 - val mse: 4.8154e-04
Epoch 113/200
e: 0.0043 - val loss: 4.5341e-04 - val mae: 0.0157 - val mse: 4.5341e-04
Epoch 114/200
16/16 [==================== ] - 1s 40ms/step - loss: 0.0041 - mae: 0.0417 - ms
e: 0.0041 - val loss: 4.5987e-04 - val mae: 0.0159 - val mse: 4.5987e-04
Epoch 115/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0043 - mae: 0.0422 - ms
e: 0.0043 - val loss: 4.5448e-04 - val mae: 0.0158 - val mse: 4.5448e-04
Epoch 116/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0039 - mae: 0.0412 - ms
e: 0.0039 - val loss: 4.6039e-04 - val mae: 0.0158 - val mse: 4.6039e-04
Epoch 117/200
16/16 [==================== ] - 1s 40ms/step - loss: 0.0042 - mae: 0.0417 - ms
e: 0.0042 - val loss: 4.6454e-04 - val mae: 0.0158 - val mse: 4.6454e-04
Epoch 118/200
e: 0.0042 - val loss: 4.4896e-04 - val mae: 0.0156 - val mse: 4.4896e-04
Epoch 119/200
e: 0.0041 - val loss: 4.5845e-04 - val mae: 0.0157 - val mse: 4.5845e-04
Epoch 120/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0043 - mae: 0.0419 - ms
e: 0.0043 - val loss: 4.5144e-04 - val mae: 0.0156 - val mse: 4.5144e-04
Epoch 121/200
```

e: 0.0041 - val loss: 4.3475e-04 - val mae: 0.0155 - val mse: 4.3475e-04

```
Epoch 122/200
e: 0.0039 - val loss: 4.3185e-04 - val mae: 0.0152 - val mse: 4.3185e-04
Epoch 123/200
e: 0.0040 - val loss: 4.4098e-04 - val mae: 0.0155 - val mse: 4.4098e-04
e: 0.0039 - val loss: 4.3314e-04 - val mae: 0.0153 - val mse: 4.3314e-04
Epoch 125/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0040 - mae: 0.0407 - ms
e: 0.0040 - val loss: 4.3228e-04 - val mae: 0.0152 - val mse: 4.3228e-04
Epoch 126/200
16/16 [============= ] - 1s 39ms/step - loss: 0.0039 - mae: 0.0405 - ms
e: 0.0039 - val loss: 4.2755e-04 - val mae: 0.0152 - val mse: 4.2755e-04
Epoch 127/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0044 - mae: 0.0418 - ms
e: 0.0044 - val loss: 4.2712e-04 - val mae: 0.0153 - val mse: 4.2712e-04
Epoch 128/200
e: 0.0042 - val loss: 4.3306e-04 - val mae: 0.0153 - val mse: 4.3306e-04
Epoch 129/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0039 - mae: 0.0406 - ms
e: 0.0039 - val loss: 4.1441e-04 - val mae: 0.0149 - val mse: 4.1441e-04
Epoch 130/200
e: 0.0039 - val loss: 4.1673e-04 - val mae: 0.0150 - val mse: 4.1673e-04
Epoch 131/200
e: 0.0039 - val loss: 4.1578e-04 - val mae: 0.0150 - val mse: 4.1578e-04
Epoch 132/200
e: 0.0040 - val loss: 4.1821e-04 - val mae: 0.0150 - val mse: 4.1821e-04
Epoch 133/200
e: 0.0038 - val loss: 4.2313e-04 - val mae: 0.0151 - val mse: 4.2313e-04
Epoch 134/200
e: 0.0038 - val loss: 4.1312e-04 - val mae: 0.0149 - val mse: 4.1312e-04
Epoch 135/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0040 - mae: 0.0405 - ms
e: 0.0040 - val loss: 4.3038e-04 - val mae: 0.0153 - val mse: 4.3038e-04
Epoch 136/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0040 - mae: 0.0406 - ms
e: 0.0040 - val loss: 4.2680e-04 - val mae: 0.0152 - val mse: 4.2680e-04
Epoch 137/200
16/16 [============= ] - 1s 39ms/step - loss: 0.0038 - mae: 0.0400 - ms
e: 0.0038 - val loss: 4.2148e-04 - val mae: 0.0152 - val mse: 4.2148e-04
Epoch 138/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0040 - mae: 0.0405 - ms
e: 0.0040 - val loss: 4.0526e-04 - val mae: 0.0148 - val mse: 4.0526e-04
Epoch 139/200
e: 0.0039 - val loss: 4.0082e-04 - val mae: 0.0147 - val mse: 4.0082e-04
Epoch 140/200
e: 0.0038 - val loss: 4.0953e-04 - val_mae: 0.0149 - val_mse: 4.0953e-04
Epoch 141/200
e: 0.0040 - val loss: 4.0911e-04 - val mae: 0.0149 - val mse: 4.0911e-04
Epoch 142/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0038 - mae: 0.0397 - ms
e: 0.0038 - val loss: 3.9823e-04 - val mae: 0.0146 - val mse: 3.9823e-04
Epoch 143/200
```

e: 0.0037 - val loss: 3.9698e-04 - val mae: 0.0148 - val mse: 3.9698e-04

```
Epoch 144/200
e: 0.0041 - val loss: 3.9362e-04 - val mae: 0.0146 - val mse: 3.9362e-04
Epoch 145/200
e: 0.0042 - val loss: 4.0671e-04 - val mae: 0.0149 - val mse: 4.0671e-04
Epoch 146/200
e: 0.0040 - val loss: 4.2407e-04 - val mae: 0.0151 - val mse: 4.2407e-04
Epoch 147/200
e: 0.0041 - val loss: 3.8897e-04 - val mae: 0.0145 - val mse: 3.8897e-04
Epoch 148/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0039 - mae: 0.0396 - ms
e: 0.0039 - val loss: 3.8152e-04 - val mae: 0.0143 - val mse: 3.8152e-04
Epoch 149/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0040 - mae: 0.0399 - ms
e: 0.0040 - val loss: 4.0871e-04 - val mae: 0.0148 - val mse: 4.0871e-04
Epoch 150/200
e: 0.0038 - val loss: 3.8049e-04 - val mae: 0.0143 - val mse: 3.8049e-04
Epoch 151/200
16/16 [=============== ] - 1s 44ms/step - loss: 0.0040 - mae: 0.0397 - ms
e: 0.0040 - val loss: 3.9043e-04 - val mae: 0.0146 - val mse: 3.9043e-04
Epoch 152/200
e: 0.0039 - val loss: 3.9486e-04 - val mae: 0.0146 - val mse: 3.9486e-04
Epoch 153/200
e: 0.0037 - val loss: 3.8918e-04 - val mae: 0.0146 - val mse: 3.8918e-04
Epoch 154/200
e: 0.0038 - val loss: 3.7201e-04 - val mae: 0.0142 - val mse: 3.7201e-04
Epoch 155/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0392 - ms
e: 0.0038 - val loss: 3.8281e-04 - val mae: 0.0144 - val mse: 3.8281e-04
Epoch 156/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0038 - mae: 0.0392 - ms
e: 0.0038 - val loss: 3.7056e-04 - val mae: 0.0141 - val mse: 3.7056e-04
Epoch 157/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0038 - mae: 0.0392 - ms
e: 0.0038 - val loss: 3.9410e-04 - val mae: 0.0146 - val mse: 3.9410e-04
Epoch 158/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0039 - mae: 0.0394 - ms
e: 0.0039 - val loss: 3.8401e-04 - val mae: 0.0145 - val mse: 3.8401e-04
Epoch 159/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0390 - ms
e: 0.0038 - val loss: 3.6975e-04 - val mae: 0.0141 - val mse: 3.6975e-04
Epoch 160/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0040 - mae: 0.0396 - ms
e: 0.0040 - val loss: 3.6165e-04 - val mae: 0.0141 - val mse: 3.6165e-04
Epoch 161/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0038 - mae: 0.0390 - ms
e: 0.0038 - val loss: 3.7688e-04 - val mae: 0.0144 - val mse: 3.7688e-04
Epoch 162/200
e: 0.0038 - val loss: 3.9536e-04 - val mae: 0.0146 - val mse: 3.9536e-04
Epoch 163/200
e: 0.0039 - val loss: 3.8389e-04 - val mae: 0.0144 - val mse: 3.8389e-04
Epoch 164/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0040 - mae: 0.0394 - ms
e: 0.0040 - val loss: 3.7857e-04 - val mae: 0.0143 - val mse: 3.7857e-04
Epoch 165/200
```

e: 0.0037 - val loss: 3.5651e-04 - val mae: 0.0139 - val mse: 3.5651e-04

```
Epoch 166/200
e: 0.0039 - val loss: 3.6982e-04 - val mae: 0.0141 - val mse: 3.6982e-04
Epoch 167/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0039 - mae: 0.0391 - ms
e: 0.0039 - val loss: 3.7766e-04 - val mae: 0.0145 - val mse: 3.7766e-04
Epoch 168/200
e: 0.0036 - val loss: 3.5829e-04 - val mae: 0.0139 - val mse: 3.5829e-04
Epoch 169/200
e: 0.0039 - val loss: 3.6265e-04 - val mae: 0.0140 - val mse: 3.6265e-04
Epoch 170/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0040 - mae: 0.0391 - ms
e: 0.0040 - val loss: 3.5820e-04 - val mae: 0.0140 - val mse: 3.5820e-04
Epoch 171/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0039 - mae: 0.0390 - ms
e: 0.0039 - val loss: 3.6691e-04 - val mae: 0.0141 - val mse: 3.6691e-04
Epoch 172/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0387 - ms
e: 0.0038 - val loss: 3.7628e-04 - val mae: 0.0143 - val mse: 3.7628e-04
Epoch 173/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0038 - mae: 0.0389 - ms
e: 0.0038 - val loss: 3.5734e-04 - val mae: 0.0139 - val mse: 3.5734e-04
Epoch 174/200
e: 0.0038 - val loss: 3.5051e-04 - val mae: 0.0138 - val mse: 3.5051e-04
Epoch 175/200
e: 0.0037 - val loss: 3.5973e-04 - val mae: 0.0139 - val mse: 3.5973e-04
Epoch 176/200
e: 0.0037 - val loss: 3.5883e-04 - val mae: 0.0139 - val mse: 3.5883e-04
Epoch 177/200
e: 0.0038 - val loss: 3.5293e-04 - val mae: 0.0138 - val mse: 3.5293e-04
Epoch 178/200
e: 0.0038 - val loss: 3.5356e-04 - val mae: 0.0138 - val mse: 3.5356e-04
Epoch 179/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0384 - ms
e: 0.0038 - val loss: 3.6419e-04 - val mae: 0.0140 - val mse: 3.6419e-04
Epoch 180/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0036 - mae: 0.0380 - ms
e: 0.0036 - val loss: 3.5491e-04 - val mae: 0.0138 - val mse: 3.5491e-04
Epoch 181/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0038 - mae: 0.0382 - ms
e: 0.0038 - val loss: 3.5188e-04 - val mae: 0.0138 - val mse: 3.5188e-04
Epoch 182/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0387 - ms
e: 0.0038 - val loss: 3.6648e-04 - val mae: 0.0141 - val mse: 3.6648e-04
Epoch 183/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0038 - mae: 0.0384 - ms
e: 0.0038 - val loss: 3.3706e-04 - val mae: 0.0136 - val mse: 3.3706e-04
Epoch 184/200
e: 0.0037 - val loss: 3.5051e-04 - val mae: 0.0137 - val mse: 3.5051e-04
Epoch 185/200
e: 0.0038 - val loss: 3.3681e-04 - val mae: 0.0134 - val mse: 3.3681e-04
Epoch 186/200
16/16 [=============== ] - 1s 39ms/step - loss: 0.0036 - mae: 0.0377 - ms
e: 0.0036 - val loss: 4.0160e-04 - val mae: 0.0149 - val mse: 4.0160e-04
Epoch 187/200
```

e: 0.0037 - val loss: 3.4327e-04 - val mae: 0.0136 - val mse: 3.4327e-04

```
e: 0.0037 - val loss: 3.4056e-04 - val mae: 0.0136 - val mse: 3.4056e-04
Epoch 189/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0039 - mae: 0.0386 - ms
e: 0.0039 - val loss: 3.3286e-04 - val mae: 0.0135 - val mse: 3.3286e-04
Epoch 190/200
e: 0.0038 - val loss: 3.4527e-04 - val mae: 0.0136 - val mse: 3.4527e-04
Epoch 191/200
e: 0.0039 - val loss: 3.5288e-04 - val mae: 0.0141 - val mse: 3.5288e-04
Epoch 192/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0039 - mae: 0.0382 - ms
e: 0.0039 - val loss: 3.2529e-04 - val mae: 0.0132 - val mse: 3.2529e-04
Epoch 193/200
e: 0.0037 - val loss: 3.3470e-04 - val mae: 0.0134 - val mse: 3.3470e-04
Epoch 194/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0036 - mae: 0.0377 - ms
e: 0.0036 - val loss: 3.3858e-04 - val mae: 0.0136 - val mse: 3.3858e-04
Epoch 195/200
16/16 [============== ] - 1s 39ms/step - loss: 0.0038 - mae: 0.0378 - ms
e: 0.0038 - val loss: 3.3438e-04 - val mae: 0.0134 - val mse: 3.3438e-04
Epoch 196/200
e: 0.0039 - val loss: 3.4048e-04 - val mae: 0.0135 - val mse: 3.4048e-04
Epoch 197/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0036 - mae: 0.0379 - ms
e: 0.0036 - val loss: 3.5022e-04 - val mae: 0.0138 - val mse: 3.5022e-04
Epoch 198/200
e: 0.0039 - val loss: 3.3582e-04 - val mae: 0.0134 - val mse: 3.3582e-04
Epoch 199/200
e: 0.0036 - val loss: 3.4151e-04 - val mae: 0.0135 - val mse: 3.4151e-04
Epoch 200/200
e: 0.0037 - val loss: 3.5051e-04 - val mae: 0.0139 - val mse: 3.5051e-04
Epoch 1/200
2023-05-09 18:22:07.890320: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.159680: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.254535: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.339405: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.424328: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.558784: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.693577: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:08.846095: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
1/16 [>.....] - ETA: 46s - loss: 0.0146 - mae: 0.0919 - mse: 0.
0146
2023-05-09 18:22:08.999646: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
134
2023-05-09 18:22:10.727939: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
```

2023-05-09 18:22:10.813257: I tensorflow/core/grappler/optimizers/custom_graph_optimizer

Epoch 188/200

```
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:10.883434: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
e: 0.0134 - val loss: 0.0028 - val mae: 0.0411 - val mse: 0.0028
Epoch 2/200
2023-05-09 18:22:10.955338: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:22:11.031929: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
16/16 [============== ] - 1s 46ms/step - loss: 0.0115 - mae: 0.0810 - ms
e: 0.0115 - val loss: 0.0024 - val mae: 0.0379 - val mse: 0.0024
Epoch 3/200
e: 0.0100 - val loss: 0.0020 - val mae: 0.0347 - val mse: 0.0020
Epoch 4/200
e: 0.0086 - val loss: 0.0017 - val mae: 0.0316 - val mse: 0.0017
e: 0.0074 - val loss: 0.0015 - val mae: 0.0290 - val mse: 0.0015
Epoch 6/200
e: 0.0067 - val loss: 0.0013 - val mae: 0.0274 - val mse: 0.0013
Epoch 7/200
e: 0.0062 - val loss: 0.0012 - val mae: 0.0270 - val mse: 0.0012
Epoch 8/200
e: 0.0060 - val loss: 0.0012 - val mae: 0.0270 - val mse: 0.0012
Epoch 9/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0061 - mae: 0.0560 - ms
e: 0.0061 - val loss: 0.0012 - val mae: 0.0270 - val mse: 0.0012
Epoch 10/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0061 - mae: 0.0559 - ms
e: 0.0061 - val loss: 0.0012 - val_mae: 0.0270 - val_mse: 0.0012
Epoch 11/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0060 - mae: 0.0557 - ms
e: 0.0060 - val loss: 0.0013 - val mae: 0.0273 - val mse: 0.0013
Epoch 12/200
e: 0.0061 - val loss: 0.0012 - val mae: 0.0271 - val mse: 0.0012
Epoch 13/200
e: 0.0060 - val loss: 0.0013 - val mae: 0.0272 - val mse: 0.0013
Epoch 14/200
e: 0.0062 - val loss: 0.0013 - val mae: 0.0273 - val mse: 0.0013
Epoch 15/200
16/16 [==================== ] - 1s 44ms/step - loss: 0.0062 - mae: 0.0564 - ms
e: 0.0062 - val loss: 0.0013 - val mae: 0.0273 - val mse: 0.0013
Epoch 16/200
e: 0.0062 - val loss: 0.0013 - val mae: 0.0275 - val mse: 0.0013
Epoch 17/200
e: 0.0063 - val loss: 0.0013 - val mae: 0.0277 - val mse: 0.0013
Epoch 18/200
e: 0.0062 - val loss: 0.0013 - val mae: 0.0277 - val mse: 0.0013
Epoch 19/200
16/16 [=================== ] - 1s 44ms/step - loss: 0.0061 - mae: 0.0571 - ms
e: 0.0061 - val loss: 0.0013 - val mae: 0.0278 - val mse: 0.0013
Epoch 20/200
```

```
e: 0.0063 - val loss: 0.0013 - val_mae: 0.0281 - val_mse: 0.0013
Epoch 21/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0062 - mae: 0.0570 - ms
e: 0.0062 - val loss: 0.0013 - val mae: 0.0281 - val mse: 0.0013
Epoch 22/200
e: 0.0063 - val loss: 0.0014 - val mae: 0.0283 - val mse: 0.0014
Epoch 23/200
e: 0.0067 - val loss: 0.0014 - val mae: 0.0289 - val mse: 0.0014
Epoch 24/200
e: 0.0065 - val loss: 0.0014 - val mae: 0.0287 - val mse: 0.0014
Epoch 25/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0065 - mae: 0.0584 - ms
e: 0.0065 - val loss: 0.0014 - val mae: 0.0288 - val mse: 0.0014
Epoch 26/200
e: 0.0066 - val loss: 0.0014 - val mae: 0.0289 - val mse: 0.0014
Epoch 27/200
e: 0.0070 - val loss: 0.0014 - val mae: 0.0292 - val mse: 0.0014
Epoch 28/200
e: 0.0067 - val loss: 0.0014 - val mae: 0.0292 - val mse: 0.0014
Epoch 29/200
16/16 [=============== ] - 1s 46ms/step - loss: 0.0067 - mae: 0.0598 - ms
e: 0.0067 - val loss: 0.0015 - val mae: 0.0294 - val mse: 0.0015
Epoch 30/200
e: 0.0069 - val loss: 0.0015 - val mae: 0.0296 - val mse: 0.0015
Epoch 31/200
e: 0.0068 - val loss: 0.0015 - val mae: 0.0297 - val mse: 0.0015
Epoch 32/200
16/16 [============== ] - 1s 44ms/step - loss: 0.0071 - mae: 0.0616 - ms
e: 0.0071 - val loss: 0.0015 - val mae: 0.0296 - val mse: 0.0015
Epoch 33/200
e: 0.0072 - val loss: 0.0015 - val mae: 0.0298 - val mse: 0.0015
Epoch 34/200
e: 0.0071 - val loss: 0.0015 - val mae: 0.0300 - val mse: 0.0015
Epoch 35/200
e: 0.0072 - val loss: 0.0015 - val mae: 0.0300 - val mse: 0.0015
Epoch 36/200
e: 0.0073 - val loss: 0.0015 - val mae: 0.0300 - val mse: 0.0015
Epoch 37/200
16/16 [=============== ] - 1s 45ms/step - loss: 0.0074 - mae: 0.0627 - ms
e: 0.0074 - val loss: 0.0015 - val mae: 0.0301 - val mse: 0.0015
Epoch 38/200
e: 0.0075 - val loss: 0.0015 - val mae: 0.0302 - val mse: 0.0015
Epoch 39/200
e: 0.0075 - val loss: 0.0015 - val mae: 0.0302 - val mse: 0.0015
Epoch 40/200
e: 0.0075 - val loss: 0.0016 - val mae: 0.0304 - val mse: 0.0016
Epoch 41/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0075 - mae: 0.0636 - ms
e: 0.0075 - val loss: 0.0016 - val mae: 0.0313 - val mse: 0.0016
Epoch 42/200
```

16/16 [===============] - 1s 45ms/step - loss: 0.0076 - mae: 0.0638 - ms

```
e: 0.0076 - val loss: 0.0015 - val_mae: 0.0303 - val_mse: 0.0015
Epoch 43/200
16/16 [============== ] - 1s 44ms/step - loss: 0.0073 - mae: 0.0628 - ms
e: 0.0073 - val loss: 0.0015 - val mae: 0.0303 - val mse: 0.0015
Epoch 44/200
e: 0.0074 - val loss: 0.0015 - val mae: 0.0303 - val mse: 0.0015
Epoch 45/200
e: 0.0075 - val loss: 0.0015 - val mae: 0.0304 - val mse: 0.0015
Epoch 46/200
e: 0.0074 - val loss: 0.0015 - val mae: 0.0305 - val mse: 0.0015
Epoch 47/200
16/16 [============= ] - 1s 43ms/step - loss: 0.0073 - mae: 0.0629 - ms
e: 0.0073 - val loss: 0.0015 - val mae: 0.0301 - val mse: 0.0015
Epoch 48/200
e: 0.0072 - val loss: 0.0015 - val mae: 0.0303 - val mse: 0.0015
Epoch 49/200
e: 0.0073 - val loss: 0.0015 - val mae: 0.0303 - val mse: 0.0015
Epoch 50/200
e: 0.0074 - val loss: 0.0015 - val mae: 0.0300 - val mse: 0.0015
Epoch 51/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0072 - mae: 0.0630 - ms
e: 0.0072 - val loss: 0.0015 - val mae: 0.0299 - val mse: 0.0015
Epoch 52/200
e: 0.0072 - val loss: 0.0015 - val mae: 0.0302 - val mse: 0.0015
Epoch 53/200
e: 0.0073 - val loss: 0.0015 - val mae: 0.0298 - val mse: 0.0015
Epoch 54/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0074 - mae: 0.0629 - ms
e: 0.0074 - val loss: 0.0015 - val mae: 0.0296 - val mse: 0.0015
Epoch 55/200
e: 0.0072 - val loss: 0.0015 - val mae: 0.0295 - val mse: 0.0015
Epoch 56/200
e: 0.0071 - val loss: 0.0014 - val mae: 0.0293 - val mse: 0.0014
Epoch 57/200
16/16 [==================== ] - 1s 41ms/step - loss: 0.0071 - mae: 0.0616 - ms
e: 0.0071 - val loss: 0.0014 - val mae: 0.0290 - val mse: 0.0014
Epoch 58/200
e: 0.0070 - val loss: 0.0014 - val mae: 0.0295 - val mse: 0.0014
Epoch 59/200
e: 0.0071 - val loss: 0.0014 - val mae: 0.0288 - val mse: 0.0014
Epoch 60/200
e: 0.0069 - val loss: 0.0014 - val mae: 0.0287 - val mse: 0.0014
Epoch 61/200
e: 0.0068 - val loss: 0.0013 - val mae: 0.0281 - val mse: 0.0013
Epoch 62/200
e: 0.0065 - val loss: 0.0013 - val mae: 0.0279 - val mse: 0.0013
Epoch 63/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0068 - mae: 0.0597 - ms
e: 0.0068 - val loss: 0.0013 - val mae: 0.0276 - val mse: 0.0013
Epoch 64/200
```

16/16 [===============] - 1s 40ms/step - loss: 0.0066 - mae: 0.0589 - ms

```
e: 0.0066 - val loss: 0.0013 - val_mae: 0.0274 - val_mse: 0.0013
Epoch 65/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0063 - mae: 0.0581 - ms
e: 0.0063 - val loss: 0.0013 - val mae: 0.0271 - val mse: 0.0013
Epoch 66/200
e: 0.0063 - val loss: 0.0012 - val mae: 0.0267 - val mse: 0.0012
Epoch 67/200
e: 0.0063 - val loss: 0.0012 - val mae: 0.0264 - val mse: 0.0012
Epoch 68/200
e: 0.0061 - val loss: 0.0012 - val mae: 0.0259 - val mse: 0.0012
e: 0.0060 - val loss: 0.0011 - val mae: 0.0256 - val mse: 0.0011
Epoch 70/200
e: 0.0060 - val loss: 0.0011 - val mae: 0.0251 - val mse: 0.0011
Epoch 71/200
e: 0.0059 - val loss: 0.0011 - val mae: 0.0248 - val mse: 0.0011
Epoch 72/200
e: 0.0059 - val loss: 0.0010 - val mae: 0.0244 - val mse: 0.0010
Epoch 73/200
e: 0.0057 - val loss: 0.0010 - val mae: 0.0240 - val mse: 0.0010
Epoch 74/200
e: 0.0056 - val loss: 9.8473e-04 - val mae: 0.0236 - val mse: 9.8473e-04
Epoch 75/200
e: 0.0056 - val loss: 9.6353e-04 - val mae: 0.0233 - val mse: 9.6353e-04
Epoch 76/200
16/16 [============== ] - 1s 44ms/step - loss: 0.0054 - mae: 0.0526 - ms
e: 0.0054 - val loss: 9.3353e-04 - val mae: 0.0229 - val mse: 9.3353e-04
Epoch 77/200
e: 0.0054 - val loss: 9.0779e-04 - val mae: 0.0226 - val mse: 9.0779e-04
Epoch 78/200
e: 0.0054 - val loss: 8.9245e-04 - val mae: 0.0224 - val mse: 8.9245e-04
Epoch 79/200
e: 0.0056 - val loss: 8.5858e-04 - val mae: 0.0219 - val mse: 8.5858e-04
Epoch 80/200
e: 0.0053 - val loss: 8.6052e-04 - val mae: 0.0218 - val mse: 8.6052e-04
Epoch 81/200
e: 0.0053 - val loss: 8.1474e-04 - val mae: 0.0213 - val mse: 8.1474e-04
Epoch 82/200
e: 0.0051 - val loss: 7.9858e-04 - val mae: 0.0211 - val mse: 7.9858e-04
Epoch 83/200
16/16 [==================== ] - 1s 41ms/step - loss: 0.0050 - mae: 0.0496 - ms
e: 0.0050 - val loss: 7.7896e-04 - val mae: 0.0208 - val mse: 7.7896e-04
Epoch 84/200
e: 0.0051 - val loss: 7.9226e-04 - val mae: 0.0209 - val mse: 7.9226e-04
Epoch 85/200
e: 0.0051 - val loss: 7.9997e-04 - val mae: 0.0211 - val mse: 7.9997e-04
Epoch 86/200
```

```
e: 0.0055 - val loss: 7.7131e-04 - val mae: 0.0208 - val mse: 7.7131e-04
Epoch 87/200
16/16 [============== ] - 1s 45ms/step - loss: 0.0054 - mae: 0.0501 - ms
e: 0.0054 - val loss: 7.6371e-04 - val mae: 0.0207 - val mse: 7.6371e-04
Epoch 88/200
e: 0.0048 - val loss: 7.5996e-04 - val mae: 0.0206 - val mse: 7.5996e-04
Epoch 89/200
e: 0.0048 - val loss: 7.7082e-04 - val mae: 0.0208 - val mse: 7.7082e-04
Epoch 90/200
e: 0.0051 - val loss: 7.7719e-04 - val mae: 0.0209 - val mse: 7.7719e-04
Epoch 91/200
e: 0.0049 - val loss: 7.6215e-04 - val mae: 0.0206 - val mse: 7.6215e-04
Epoch 92/200
e: 0.0049 - val loss: 7.6698e-04 - val mae: 0.0207 - val mse: 7.6698e-04
Epoch 93/200
e: 0.0052 - val loss: 7.5454e-04 - val mae: 0.0207 - val mse: 7.5454e-04
Epoch 94/200
e: 0.0053 - val loss: 7.6844e-04 - val mae: 0.0207 - val mse: 7.6844e-04
Epoch 95/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0049 - mae: 0.0484 - ms
e: 0.0049 - val loss: 7.8014e-04 - val mae: 0.0209 - val mse: 7.8014e-04
Epoch 96/200
e: 0.0051 - val loss: 7.8461e-04 - val mae: 0.0209 - val mse: 7.8461e-04
Epoch 97/200
e: 0.0050 - val loss: 7.9508e-04 - val mae: 0.0213 - val mse: 7.9508e-04
Epoch 98/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0049 - mae: 0.0481 - ms
e: 0.0049 - val loss: 7.5906e-04 - val mae: 0.0206 - val mse: 7.5906e-04
Epoch 99/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0049 - mae: 0.0489 - ms
e: 0.0049 - val loss: 7.5795e-04 - val mae: 0.0206 - val mse: 7.5795e-04
Epoch 100/200
e: 0.0052 - val loss: 7.6526e-04 - val mae: 0.0207 - val mse: 7.6526e-04
Epoch 101/200
e: 0.0050 - val loss: 7.4885e-04 - val mae: 0.0205 - val mse: 7.4885e-04
Epoch 102/200
16/16 [=============== ] - 1s 44ms/step - loss: 0.0050 - mae: 0.0490 - ms
e: 0.0050 - val loss: 7.7252e-04 - val mae: 0.0208 - val mse: 7.7252e-04
Epoch 103/200
e: 0.0050 - val_loss: 7.6310e-04 - val_mae: 0.0206 - val_mse: 7.6310e-04
Epoch 104/200
e: 0.0048 - val loss: 7.8179e-04 - val mae: 0.0211 - val mse: 7.8179e-04
Epoch 105/200
e: 0.0047 - val loss: 7.6483e-04 - val mae: 0.0207 - val mse: 7.6483e-04
Epoch 106/200
e: 0.0050 - val loss: 7.6217e-04 - val mae: 0.0207 - val mse: 7.6217e-04
Epoch 107/200
e: 0.0049 - val loss: 7.5007e-04 - val mae: 0.0205 - val mse: 7.5007e-04
Epoch 108/200
```

```
e: 0.0047 - val loss: 7.6349e-04 - val mae: 0.0207 - val mse: 7.6349e-04
Epoch 109/200
16/16 [============== ] - 1s 42ms/step - loss: 0.0051 - mae: 0.0488 - ms
e: 0.0051 - val loss: 7.7402e-04 - val mae: 0.0208 - val mse: 7.7402e-04
Epoch 110/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0051 - mae: 0.0491 - ms
e: 0.0051 - val loss: 7.6060e-04 - val mae: 0.0206 - val mse: 7.6060e-04
Epoch 111/200
e: 0.0051 - val loss: 7.5548e-04 - val mae: 0.0206 - val mse: 7.5548e-04
Epoch 112/200
e: 0.0050 - val loss: 7.6999e-04 - val mae: 0.0209 - val mse: 7.6999e-04
e: 0.0052 - val loss: 7.8769e-04 - val mae: 0.0211 - val mse: 7.8769e-04
Epoch 114/200
e: 0.0049 - val_loss: 7.4975e-04 - val_mae: 0.0205 - val_mse: 7.4975e-04
Epoch 115/200
e: 0.0050 - val loss: 7.6021e-04 - val mae: 0.0206 - val mse: 7.6021e-04
Epoch 116/200
e: 0.0049 - val loss: 7.4371e-04 - val mae: 0.0205 - val mse: 7.4371e-04
Epoch 117/200
e: 0.0050 - val loss: 7.4051e-04 - val mae: 0.0204 - val mse: 7.4051e-04
Epoch 118/200
e: 0.0048 - val loss: 7.3505e-04 - val mae: 0.0204 - val mse: 7.3505e-04
Epoch 119/200
e: 0.0052 - val loss: 7.4252e-04 - val mae: 0.0204 - val mse: 7.4252e-04
Epoch 120/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0051 - mae: 0.0488 - ms
e: 0.0051 - val loss: 7.6077e-04 - val mae: 0.0207 - val mse: 7.6077e-04
Epoch 121/200
e: 0.0050 - val loss: 7.3975e-04 - val mae: 0.0204 - val mse: 7.3975e-04
Epoch 122/200
e: 0.0050 - val loss: 7.5046e-04 - val mae: 0.0206 - val mse: 7.5046e-04
Epoch 123/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0050 - mae: 0.0484 - ms
e: 0.0050 - val loss: 7.3125e-04 - val mae: 0.0202 - val mse: 7.3125e-04
Epoch 124/200
e: 0.0049 - val loss: 7.4103e-04 - val mae: 0.0203 - val mse: 7.4103e-04
Epoch 125/200
e: 0.0049 - val_loss: 7.5188e-04 - val_mae: 0.0205 - val_mse: 7.5188e-04
Epoch 126/200
e: 0.0050 - val loss: 7.3117e-04 - val mae: 0.0202 - val mse: 7.3117e-04
Epoch 127/200
16/16 [==================== ] - 1s 41ms/step - loss: 0.0050 - mae: 0.0484 - ms
e: 0.0050 - val loss: 7.4953e-04 - val mae: 0.0204 - val mse: 7.4953e-04
Epoch 128/200
e: 0.0048 - val loss: 7.3555e-04 - val mae: 0.0204 - val mse: 7.3555e-04
Epoch 129/200
e: 0.0050 - val loss: 7.4275e-04 - val mae: 0.0205 - val mse: 7.4275e-04
Epoch 130/200
```

```
e: 0.0049 - val loss: 7.2343e-04 - val mae: 0.0201 - val mse: 7.2343e-04
Epoch 131/200
16/16 [============== ] - 1s 40ms/step - loss: 0.0049 - mae: 0.0487 - ms
e: 0.0049 - val loss: 7.4931e-04 - val mae: 0.0204 - val mse: 7.4931e-04
Epoch 132/200
16/16 [=================== ] - 1s 41ms/step - loss: 0.0049 - mae: 0.0482 - ms
e: 0.0049 - val loss: 7.4925e-04 - val mae: 0.0207 - val mse: 7.4925e-04
Epoch 133/200
e: 0.0049 - val loss: 7.3238e-04 - val mae: 0.0202 - val mse: 7.3238e-04
Epoch 134/200
e: 0.0050 - val loss: 7.3123e-04 - val mae: 0.0203 - val mse: 7.3123e-04
e: 0.0050 - val loss: 7.2965e-04 - val_mae: 0.0202 - val_mse: 7.2965e-04
Epoch 136/200
e: 0.0049 - val_loss: 7.5771e-04 - val_mae: 0.0208 - val_mse: 7.5771e-04
Epoch 137/200
e: 0.0047 - val loss: 7.4517e-04 - val mae: 0.0206 - val mse: 7.4517e-04
Epoch 138/200
e: 0.0050 - val loss: 7.6153e-04 - val mae: 0.0208 - val mse: 7.6153e-04
Epoch 139/200
16/16 [=============== ] - 1s 42ms/step - loss: 0.0050 - mae: 0.0485 - ms
e: 0.0050 - val loss: 7.2636e-04 - val mae: 0.0203 - val mse: 7.2636e-04
Epoch 140/200
e: 0.0049 - val loss: 7.2065e-04 - val mae: 0.0201 - val mse: 7.2065e-04
Epoch 141/200
e: 0.0049 - val loss: 7.3066e-04 - val mae: 0.0202 - val mse: 7.3066e-04
Epoch 142/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0049 - mae: 0.0482 - ms
e: 0.0049 - val loss: 7.2388e-04 - val mae: 0.0202 - val mse: 7.2388e-04
Epoch 143/200
e: 0.0052 - val loss: 7.1761e-04 - val mae: 0.0200 - val mse: 7.1761e-04
Epoch 144/200
e: 0.0049 - val loss: 7.1904e-04 - val mae: 0.0201 - val mse: 7.1904e-04
Epoch 145/200
e: 0.0050 - val loss: 7.0351e-04 - val mae: 0.0198 - val mse: 7.0351e-04
Epoch 146/200
16/16 [==================== ] - 1s 41ms/step - loss: 0.0049 - mae: 0.0486 - ms
e: 0.0049 - val loss: 7.1711e-04 - val mae: 0.0201 - val mse: 7.1711e-04
Epoch 147/200
e: 0.0049 - val_loss: 7.0249e-04 - val_mae: 0.0198 - val_mse: 7.0249e-04
Epoch 148/200
e: 0.0047 - val loss: 6.8814e-04 - val mae: 0.0197 - val mse: 6.8814e-04
Epoch 149/200
e: 0.0049 - val loss: 6.8521e-04 - val mae: 0.0196 - val mse: 6.8521e-04
Epoch 150/200
e: 0.0051 - val loss: 7.0208e-04 - val mae: 0.0200 - val mse: 7.0208e-04
Epoch 151/200
e: 0.0050 - val loss: 7.0968e-04 - val mae: 0.0200 - val mse: 7.0968e-04
Epoch 152/200
```

```
e: 0.0047 - val loss: 6.8023e-04 - val mae: 0.0195 - val mse: 6.8023e-04
Epoch 153/200
e: 0.0047 - val loss: 6.9174e-04 - val mae: 0.0199 - val mse: 6.9174e-04
Epoch 154/200
e: 0.0048 - val loss: 6.7397e-04 - val mae: 0.0193 - val mse: 6.7397e-04
Epoch 155/200
e: 0.0048 - val loss: 6.7311e-04 - val mae: 0.0195 - val mse: 6.7311e-04
Epoch 156/200
e: 0.0047 - val loss: 6.4565e-04 - val mae: 0.0189 - val mse: 6.4565e-04
e: 0.0048 - val loss: 6.3509e-04 - val mae: 0.0188 - val mse: 6.3509e-04
Epoch 158/200
e: 0.0047 - val_loss: 6.6497e-04 - val_mae: 0.0192 - val_mse: 6.6497e-04
Epoch 159/200
e: 0.0049 - val loss: 6.1370e-04 - val mae: 0.0184 - val mse: 6.1370e-04
e: 0.0048 - val loss: 6.1786e-04 - val mae: 0.0184 - val mse: 6.1786e-04
Epoch 161/200
e: 0.0048 - val loss: 6.0880e-04 - val mae: 0.0184 - val mse: 6.0880e-04
Epoch 162/200
e: 0.0048 - val loss: 5.9276e-04 - val mae: 0.0180 - val mse: 5.9276e-04
Epoch 163/200
e: 0.0049 - val loss: 5.9120e-04 - val mae: 0.0181 - val mse: 5.9120e-04
Epoch 164/200
16/16 [============== ] - 1s 41ms/step - loss: 0.0048 - mae: 0.0478 - ms
e: 0.0048 - val loss: 5.7537e-04 - val mae: 0.0178 - val mse: 5.7537e-04
Epoch 165/200
e: 0.0047 - val loss: 5.6398e-04 - val mae: 0.0175 - val mse: 5.6398e-04
Epoch 166/200
e: 0.0050 - val loss: 5.5439e-04 - val mae: 0.0173 - val mse: 5.5439e-04
Epoch 167/200
e: 0.0046 - val loss: 5.6850e-04 - val mae: 0.0179 - val mse: 5.6850e-04
Epoch 168/200
16/16 [==================== ] - 1s 41ms/step - loss: 0.0048 - mae: 0.0481 - ms
e: 0.0048 - val loss: 5.4716e-04 - val mae: 0.0174 - val mse: 5.4716e-04
Epoch 169/200
e: 0.0047 - val loss: 5.2309e-04 - val mae: 0.0168 - val mse: 5.2309e-04
Epoch 170/200
e: 0.0046 - val loss: 5.2418e-04 - val mae: 0.0167 - val mse: 5.2418e-04
Epoch 171/200
e: 0.0048 - val loss: 5.6044e-04 - val mae: 0.0178 - val mse: 5.6044e-04
Epoch 172/200
e: 0.0049 - val loss: 5.2321e-04 - val mae: 0.0169 - val mse: 5.2321e-04
Epoch 173/200
e: 0.0047 - val loss: 5.0845e-04 - val mae: 0.0164 - val mse: 5.0845e-04
Epoch 174/200
```

```
e: 0.0048 - val loss: 5.1269e-04 - val mae: 0.0164 - val mse: 5.1269e-04
Epoch 175/200
16/16 [============== ] - 1s 46ms/step - loss: 0.0049 - mae: 0.0482 - ms
e: 0.0049 - val loss: 4.9508e-04 - val mae: 0.0162 - val mse: 4.9508e-04
Epoch 176/200
e: 0.0046 - val loss: 5.7598e-04 - val mae: 0.0183 - val mse: 5.7598e-04
Epoch 177/200
e: 0.0046 - val loss: 5.0896e-04 - val mae: 0.0166 - val mse: 5.0896e-04
Epoch 178/200
e: 0.0047 - val loss: 5.0297e-04 - val mae: 0.0163 - val mse: 5.0297e-04
e: 0.0048 - val loss: 4.9947e-04 - val_mae: 0.0164 - val_mse: 4.9947e-04
Epoch 180/200
e: 0.0048 - val loss: 5.3818e-04 - val mae: 0.0170 - val mse: 5.3818e-04
Epoch 181/200
e: 0.0048 - val loss: 5.5608e-04 - val mae: 0.0179 - val mse: 5.5608e-04
Epoch 182/200
e: 0.0049 - val loss: 4.9256e-04 - val mae: 0.0163 - val mse: 4.9256e-04
Epoch 183/200
e: 0.0051 - val loss: 5.0714e-04 - val mae: 0.0164 - val mse: 5.0714e-04
Epoch 184/200
e: 0.0047 - val loss: 4.8675e-04 - val mae: 0.0162 - val mse: 4.8675e-04
Epoch 185/200
e: 0.0051 - val loss: 5.1350e-04 - val mae: 0.0165 - val mse: 5.1350e-04
Epoch 186/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0049 - mae: 0.0495 - ms
e: 0.0049 - val loss: 4.9000e-04 - val mae: 0.0161 - val mse: 4.9000e-04
Epoch 187/200
e: 0.0050 - val loss: 4.8720e-04 - val mae: 0.0161 - val mse: 4.8720e-04
Epoch 188/200
e: 0.0051 - val loss: 5.1187e-04 - val mae: 0.0170 - val mse: 5.1187e-04
Epoch 189/200
e: 0.0048 - val loss: 4.6892e-04 - val mae: 0.0159 - val mse: 4.6892e-04
Epoch 190/200
e: 0.0051 - val loss: 4.7418e-04 - val mae: 0.0158 - val mse: 4.7418e-04
Epoch 191/200
e: 0.0052 - val_loss: 4.6930e-04 - val_mae: 0.0161 - val_mse: 4.6930e-04
Epoch 192/200
e: 0.0054 - val loss: 4.9739e-04 - val mae: 0.0169 - val mse: 4.9739e-04
Epoch 193/200
16/16 [=============== ] - 1s 40ms/step - loss: 0.0055 - mae: 0.0534 - ms
e: 0.0055 - val loss: 4.6189e-04 - val mae: 0.0160 - val mse: 4.6189e-04
Epoch 194/200
e: 0.0058 - val loss: 4.3510e-04 - val mae: 0.0151 - val mse: 4.3510e-04
Epoch 195/200
e: 0.0056 - val loss: 4.5851e-04 - val mae: 0.0161 - val mse: 4.5851e-04
Epoch 196/200
```

```
e: 0.0061 - val loss: 4.3285e-04 - val mae: 0.0154 - val mse: 4.3285e-04
Epoch 197/200
16/16 [============== ] - 1s 43ms/step - loss: 0.0063 - mae: 0.0587 - ms
e: 0.0063 - val loss: 4.4770e-04 - val mae: 0.0153 - val mse: 4.4770e-04
Epoch 198/200
16/16 [=============== ] - 1s 41ms/step - loss: 0.0065 - mae: 0.0600 - ms
e: 0.0065 - val loss: 4.3596e-04 - val mae: 0.0156 - val mse: 4.3596e-04
Epoch 199/200
16/16 [=============== ] - 1s 43ms/step - loss: 0.0066 - mae: 0.0612 - ms
e: 0.0066 - val loss: 4.4807e-04 - val mae: 0.0154 - val mse: 4.4807e-04
Epoch 200/200
e: 0.0070 - val loss: 5.1170e-04 - val mae: 0.0175 - val mse: 5.1170e-04
2023-05-09 18:24:27.105774: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 2/200
 1/253 [.....] - ETA: 2s - loss: 0.0130 - mae: 0.0883 - mse:
0.0130
2023-05-09 18:24:30.274883: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 3/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 4/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 5/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 6/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 8/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 9/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 11/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 12/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 13/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 14/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 15/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 16/200
```

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 17/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 18/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 19/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 20/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 21/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 22/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 24/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 25/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 27/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 28/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 29/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 30/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 31/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 32/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 33/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 34/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 35/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 36/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 37/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 38/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 39/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 40/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 41/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 42/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 43/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 44/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 46/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 47/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 49/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 50/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 51/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 52/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 53/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 54/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 55/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 56/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 57/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 58/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 59/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 60/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 61/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 62/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 63/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 64/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 65/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 66/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 68/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 69/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 70/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 71/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 72/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 73/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 74/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 75/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 76/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 77/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 78/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 79/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 80/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 81/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 82/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 83/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 84/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 85/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 86/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 87/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 88/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 90/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 91/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 93/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 94/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 95/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 96/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 97/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 98/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 99/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 100/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 101/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 102/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 103/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 104/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 105/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 106/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 107/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 108/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 109/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 110/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 112/200
253/253 [============ ] - 3s 10ms/step - loss: 0.0142 - mae: 0.0902 - m
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 113/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 115/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 116/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 117/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 118/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 119/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 120/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 121/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 122/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 123/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 124/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 125/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 126/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 127/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 128/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 129/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 130/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 131/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 132/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 134/200
253/253 [============ ] - 3s 10ms/step - loss: 0.0142 - mae: 0.0902 - m
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 135/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 137/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 138/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 139/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 140/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 141/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 142/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 143/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 144/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 145/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 146/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 147/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 148/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 149/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 150/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 151/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 152/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 153/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 154/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 156/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 157/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 159/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 160/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 161/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 162/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 163/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 164/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 165/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 166/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 167/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 168/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 169/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 170/200

```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 171/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 172/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 173/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 174/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 175/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 176/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 178/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 179/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 181/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 182/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 183/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 184/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 185/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 186/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 187/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 188/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 189/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 190/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 191/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
```

Epoch 192/200

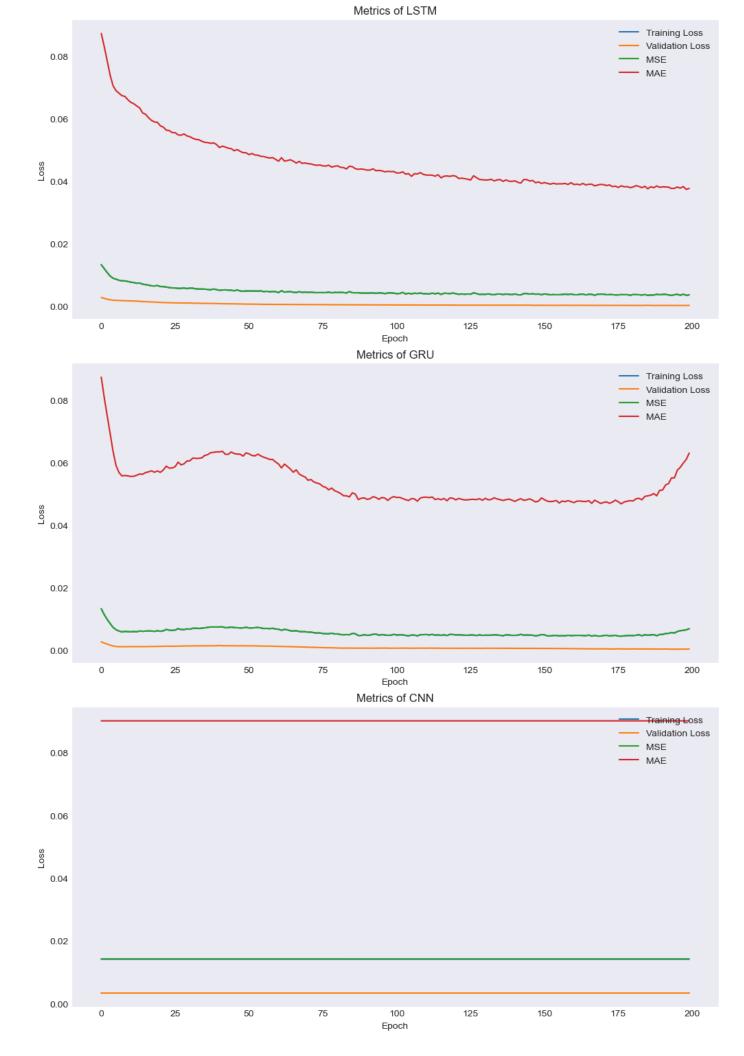
```
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 193/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 194/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 195/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 196/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 197/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 198/200
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
e: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
Epoch 200/200
se: 0.0142 - val loss: 0.0035 - val mae: 0.0453 - val mse: 0.0035
2023-05-09 18:32:24.332344: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:24.421896: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:24.494004: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
1/67 [.....] - ETA: 45s
2023-05-09 18:32:24.579274: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:24.663975: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
67/67 [======== ] - 2s 16ms/step
2023-05-09 18:32:26.192911: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:26.273908: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:26.344204: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
1/67 [.....] - ETA: 49s
2023-05-09 18:32:26.426423: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
2023-05-09 18:32:26.506756: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
67/67 [========] - 2s 16ms/step
55/67 [=============>....] - ETA: Os
2023-05-09 18:32:27.703537: I tensorflow/core/grappler/optimizers/custom graph optimizer
registry.cc:114] Plugin optimizer for device type GPU is enabled.
67/67 [=======] - 0s 3ms/step
WARNING:absl:Found untraced functions such as 1stm cell layer call fn, 1stm cell layer c
all and return conditional losses, 1stm cell 1 layer call fn, 1stm cell 1 layer call and
return conditional losses, 1stm cell 2 layer call fn while saving (showing 5 of 8). The
se functions will not be directly callable after loading.
INFO:tensorflow:Assets written to: ram://7e360b68-b942-4d3f-8f45-f07f955817ec/assets
INFO:tensorflow:Assets written to: ram://7e360b68-b942-4d3f-8f45-f07f955817ec/assets
WARNING:absl:Found untraced functions such as gru cell layer call fn, gru cell layer cal
1 and return conditional losses, gru cell 1 layer call fn, gru cell 1 layer call and ret
```

urn_conditional_losses, gru_cell_2_layer_call_fn while saving (showing 5 of 8). These fu nctions will not be directly callable after loading.

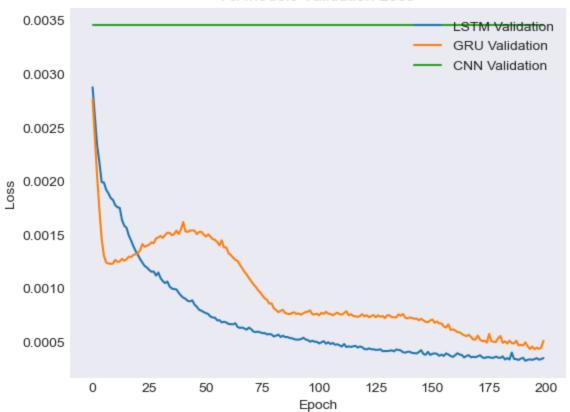
INFO:tensorflow:Assets written to: ram://0174d3c0-e587-4dbd-8e92-ae9e0739a841/assets

INFO:tensorflow:Assets written to: ram://0174d3c0-e587-4dbd-8e92-ae9e0739a841/assets ${\tt WARNING:absl:Found\ untraced\ functions\ such\ as\ _jit_compiled_convolution_op\ while\ saving}$ (showing 1 of 1). These functions will not be directly callable after loading. INFO:tensorflow:Assets written to: ram://4d2df547-0da4-4927-8da5-514dc93ad451/assets

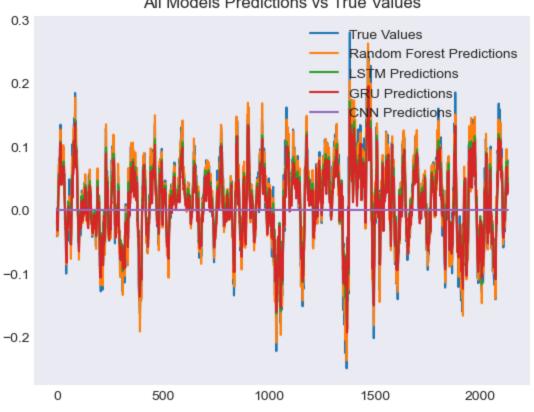
INFO:tensorflow:Assets written to: ram://4d2df547-0da4-4927-8da5-514dc93ad451/assets



All Models Validation Loss



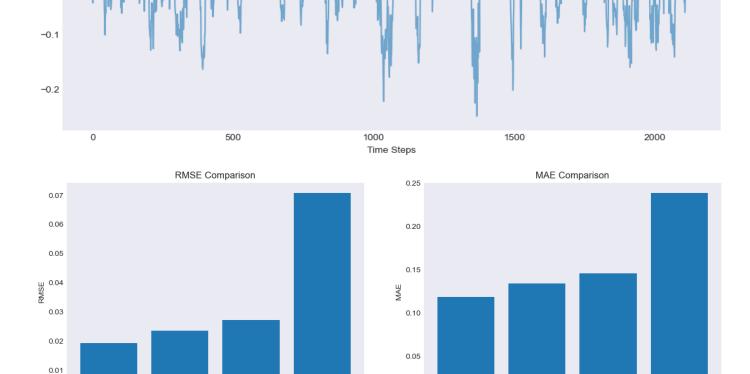
All Models Predictions vs True Values



Random Forest Predictions vs True Values







0.00

Random Forest

LSTM

GRU

Models

CNN

All Models and Root Mean Squared Error:
Random Forest's RMSE is 0.0192
LSTM's RMSE is 0.0235
GRU's RMSE is 0.0270
CNN's RMSE is 0.0707

Models

LSTM

0.00

Random Forest

The best model is Random Forest with an RMSE of 0.0192

GRU

CNN

