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## LAMPIRAN

### Lampiran 1 *Source code* model prediksi menggunakan LSTM

#### btc-1m.py

```
# %%
import datetime
import pytz
from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCBUSD"
interval = "1m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

def getDataCrypto(symbol,interval,limit,timeStart,timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+str(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
        op.append(data_json[x][1])
        hi.append(data_json[x][2])
        lo.append(data_json[x][3])
        cl.append(data_json[x][4])
        v.append(data_json[x][5])
    data = {"time": tm, "open":op,"high":hi,"low":lo,"close":cl,"volume":v}
    df = pd.DataFrame(data)
    df["time"] = df["time"].transform(lambda x: unix2utc(x))
    df["open"] = pd.to_numeric(df["open"], downcast="float")
    df["high"] = pd.to_numeric(df["high"], downcast="float")
    df["low"] = pd.to_numeric(df["low"], downcast="float")
    df["close"] = pd.to_numeric(df["close"], downcast="float")
    df["volume"] = pd.to_numeric(df["volume"], downcast="float")
    return df

df = getDataCrypto(currency,interval,len_data,None,None)
# df.drop(columns=df.columns[0], axis=1,inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
```

```

df['rsi']=rsi
df['bb_upper'] = bbands['BBL_10_2.0']
df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDS_12_26_9']

df = df.dropna()

df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential
from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)
    agg.columns = names
    # drop rows with NaN values
    if dropnan:
        agg.dropna(inplace=True)
    return agg

dataset = df
dataset = dataset.set_index('time')
# df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1
elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0,(len(dataset.columns)))]

```

```

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]
# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0], 1, values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25, input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))
model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)
# predict_data_x
predict_data_x_new = predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:, 0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new, 2)
print(time_predict_data)
print("%.2f" % inv_yhat_predict_new[0])

row_contents = [time_predict_data, inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, 'a+', newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)
        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-1m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVFYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc1m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

## btc-5m.py

```

# %%
import datetime
import pytz

from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCUSD"

```

```

interval = "5m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

def getDataCrypto(symbol, interval, limit, timeStart, timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url =
"https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+str(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
        op.append(data_json[x][1])
        hi.append(data_json[x][2])
        lo.append(data_json[x][3])
        cl.append(data_json[x][4])
        v.append(data_json[x][5])
    data = {"time": tm, "open":op,"high":hi,"low":lo,"close":cl,"volume":v}
    df = pd.DataFrame(data)
    df['time'] = df['time'].transform(lambda x: unix2utc(x))
    df["open"] = pd.to_numeric(df["open"], downcast="float")
    df["high"] = pd.to_numeric(df["high"], downcast="float")
    df["low"] = pd.to_numeric(df["low"], downcast="float")
    df["close"] = pd.to_numeric(df["close"], downcast="float")
    df["volume"] = pd.to_numeric(df["volume"], downcast="float")
    return df

df = getDataCrypto(currency,interval,len_data,None,None)
# df.drop(columns=df.columns[0], axis=1,inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
df['rsi'] = rsi
df['bb_upper'] = bbands['BBL_10_2.0']
df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDS_12_26_9']

df = df.dropna()

df = df[['time','close','volume','ma5','ma10','ma20','rsi','bb_upper','bb_lower','macd_l','macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential

```

```

from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)
    agg.columns = names
    # drop rows with NaN values
    if dropnan:
        agg.dropna(inplace=True)
    return agg

dataset = df
dataset = dataset.set_index('time')
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l1', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1
elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0,(len(dataset.columns)))]

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]
# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0],1,values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25,input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))

```



```

model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)
# predict_data_x
predict_data_x_new= predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:,0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new,2)
print(time_predict_data)
print("%.2f" %inv_yhat_predict_new[0])

row_contents = [time_predict_data,inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, 'a+', newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)
        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-5m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1t5mpGSVFYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc5m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

## btc-15m.py

```

# %%
import datetime
import pytz

from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCUSD"
interval = "15m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

def getDataCrypto(symbol,interval,limit,timeStart,timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url =
"https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+s
tr(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

```

```

# print the json response
# print(data_json)
tm,op,hi,lo,cl,v = [],[],[],[],[],[]

for x in range(len(data_json)):
# print(data_json[x][0])
tm.append(data_json[x][0])
op.append(data_json[x][1])
hi.append(data_json[x][2])
lo.append(data_json[x][3])
cl.append(data_json[x][4])
v.append(data_json[x][5])
data = {"time": tm, "open":op,"high":hi,"low":lo,"close":cl,"volume":v}
df = pd.DataFrame(data)
df['time'] = df['time'].transform(lambda x: unix2utc(x))
df["open"] = pd.to_numeric(df["open"], downcast="float")
df["high"] = pd.to_numeric(df["high"], downcast="float")
df["low"] = pd.to_numeric(df["low"], downcast="float")
df["close"] = pd.to_numeric(df["close"], downcast="float")
df["volume"] = pd.to_numeric(df["volume"], downcast="float")
return df

df = getDataCrypto(currency, interval, len_data, None, None)
# df.drop(columns=df.columns[0], axis=1, inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
df['rsi'] = rsi
df['bb_upper'] = bbands['BBL_10_2.0']
df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDS_12_26_9']

df = df.dropna()

df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential
from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)

```

```

agg.columns = names
# drop rows with NaN values
if dropnan:
    agg.dropna(inplace=True)
return agg

dataset = df
dataset = dataset.set_index('time')
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1
elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0,(len(dataset.columns)))]

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]
# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0],1,values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25,input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))
model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)
# predict_data_x
predict_data_x_new= predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:,0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new,2)
print(time_predict_data)
print("%.2f" %inv_yhat_predict_new[0])

row_contents = [time_predict_data,inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, 'a+', newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)

```

```

        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-15m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVfYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc15m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

## Lampiran 2 *Source code* model prediksi menggunakan GRU

### btc-1m.py

```

# %%
import datetime
import pytz

from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCUSD"
interval = "1m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

def getDataCrypto(symbol,interval,limit,timeStart,timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url =
"https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+s
tr(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
        op.append(data_json[x][1])
        hi.append(data_json[x][2])
        lo.append(data_json[x][3])
        cl.append(data_json[x][4])
        v.append(data_json[x][5])
    data = {"time": tm, "open":op, "high":hi, "low":lo, "close":cl, "volume":v}
    df = pd.DataFrame(data)
    df['time'] = df['time'].transform(lambda x: unix2utc(x))
    df["open"] = pd.to_numeric(df["open"], downcast="float")
    df["high"] = pd.to_numeric(df["high"], downcast="float")
    df["low"] = pd.to_numeric(df["low"], downcast="float")

```

```

df["close"] = pd.to_numeric(df["close"], downcast="float")
df["volume"] = pd.to_numeric(df["volume"], downcast="float")
return df

df = getDataCrypto(currency, interval, len_data, None, None)
# df.drop(columns=df.columns[0], axis=1, inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
df['rsi'] = rsi
df['bb_upper'] = bbands['BBL_10_2.0']
df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDS_12_26_9']

df = df.dropna()

df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential
from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)
    agg.columns = names
    # drop rows with NaN values
    if dropnan:
        agg.dropna(inplace=True)
    return agg

dataset = df
dataset = dataset.set_index('time')
# df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1

```

```

elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0,(len(dataset.columns)))]

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]
# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0],1,values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25,input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))
model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)
# predict_data_x
predict_data_x_new= predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:,0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new,2)
print(time_predict_data)
print("%.2f" %inv_yhat_predict_new[0])

row_contents = [time_predict_data,inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, 'a+', newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)
        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-1m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVFYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc1m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

```
})
```

## btc-5m.py

```
# %%
import datetime
import pytz

from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCUSD"
interval = "5m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

def getDataCrypto(symbol,interval,limit,timeStart,timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url =
"https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+
tr(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
        op.append(data_json[x][1])
        hi.append(data_json[x][2])
        lo.append(data_json[x][3])
        cl.append(data_json[x][4])
        v.append(data_json[x][5])
    data = {"time": tm, "open":op, "high":hi, "low":lo, "close":cl, "volume":v}
    df = pd.DataFrame(data)
    df['time'] = df['time'].transform(lambda x: unix2utc(x))
    df['open'] = pd.to_numeric(df["open"], downcast="float")
    df['high'] = pd.to_numeric(df["high"], downcast="float")
    df['low'] = pd.to_numeric(df["low"], downcast="float")
    df['close'] = pd.to_numeric(df["close"], downcast="float")
    df['volume'] = pd.to_numeric(df["volume"], downcast="float")
    return df

df = getDataCrypto(currency,interval,len_data,None,None)
# df.drop(columns=df.columns[0], axis=1,inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
df['rsi'] = rsi
df['bb_upper'] = bbands['BBL_10_2.0']
```

```

df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDs_12_26_9']

df = df.dropna()

df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential
from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)
    agg.columns = names
    # drop rows with NaN values
    if dropnan:
        agg.dropna(inplace=True)
    return agg

dataset = df
dataset = dataset.set_index('time')
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1
elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0,(len(dataset.columns)))]

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]

```



```

# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0],1,values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25,input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))
model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)
# predict_data_x
predict_data_x_new= predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:,0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new,2)
print(time_predict_data)
print("%.2f" %inv_yhat_predict_new[0])

row_contents = [time_predict_data,inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, "a+", newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)
        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-5m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVfYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc5m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

## btc-15m.py

```

# %%
import datetime
import pytz

from urllib.request import urlopen
import json
import pandas as pd
import pandas_ta as ta
import time
import datetime
import math

currency = "BTCUSD"
interval = "15m"
len_data = "1000"

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

```

```

def getDataCrypto(symbol, interval, limit, timeStart, timeEnd):

    # store the URL in url as
    # parameter for urlopen
    if((timeStart and timeEnd) is None):
        url = "https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit
    else:
        url =
"https://api.binance.me/api/v3/klines?symbol="+str(symbol)+"&interval="+str(interval)+"&limit="+limit+"&startTime="+s
tr(timeStart)+"&endTime="+str(timeEnd)

    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
        op.append(data_json[x][1])
        hi.append(data_json[x][2])
        lo.append(data_json[x][3])
        cl.append(data_json[x][4])
        v.append(data_json[x][5])
    data = {"time": tm, "open":op, "high":hi, "low":lo, "close":cl, "volume":v}
    df = pd.DataFrame(data)
    df['time'] = df['time'].transform(lambda x: unix2utc(x))
    df["open"] = pd.to_numeric(df["open"], downcast="float")
    df["high"] = pd.to_numeric(df["high"], downcast="float")
    df["low"] = pd.to_numeric(df["low"], downcast="float")
    df["close"] = pd.to_numeric(df["close"], downcast="float")
    df["volume"] = pd.to_numeric(df["volume"], downcast="float")
    return df

df = getDataCrypto(currency, interval, len_data, None, None)
# df.drop(columns=df.columns[0], axis=1, inplace=True)

# Calculate Indicator Technical
import pandas_ta as ta

sma5 = ta.sma(df['close'], length=5)
sma10 = ta.sma(df['close'], length=10)
sma20 = ta.sma(df['close'], length=20)
rsi = ta.rsi(df['close'].astype(float), length = 10)
macd = ta.macd(df['close'].astype(float), length = 10)
bbands = ta.bbands(df['close'].astype(float), length = 10)

df['ma5'] = sma5
df['ma10'] = sma10
df['ma20'] = sma20
df['rsi'] = rsi
df['bb_upper'] = bbands['BBL_10_2.0']
df['bb_lower'] = bbands['BBU_10_2.0']
df['bb_middle'] = bbands['BBM_10_2.0']
df['macd_l'] = macd['MACD_12_26_9']
df['macd_s'] = macd['MACDS_12_26_9']

df = df.dropna()

df = df[['time', 'close', 'volume', 'ma5', 'ma10', 'ma20', 'rsi', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s']]

from math import sqrt
from numpy import concatenate
from matplotlib import pyplot
from pandas import read_csv
from pandas import DataFrame
from pandas import concat
from sklearn.preprocessing import MinMaxScaler
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import mean_squared_error
from keras.models import Sequential
from keras.models import load_model
from keras.layers import Dense
from keras.layers import LSTM

import pandas as pd

```

```

# convert series to supervised learning
def series_to_supervised(data, n_in=1, n_out=1, dropnan=True):
    n_vars = 1 if type(data) is list else data.shape[1]
    df = DataFrame(data)
    cols, names = list(), list()
    # input sequence (t-n, ... t-1)
    for i in range(n_in, 0, -1):
        cols.append(df.shift(i))
        names += [('var%d(t-%d)' % (j+1, i)) for j in range(n_vars)]
    # forecast sequence (t, t+1, ... t+n)
    for i in range(0, n_out):
        cols.append(df.shift(-i))
        if i == 0:
            names += [('var%d(t)' % (j+1)) for j in range(n_vars)]
        else:
            names += [('var%d(t+%d)' % (j+1, i)) for j in range(n_vars)]
    # put it all together
    agg = concat(cols, axis=1)
    agg.columns = names
    # drop rows with NaN values
    if dropnan:
        agg.dropna(inplace=True)
    return agg

dataset = df
dataset = dataset.set_index('time')
# dataset = dataset.drop(columns=['ma5', 'ma10', 'ma20', 'bb_upper', 'bb_lower', 'macd_l', 'macd_s'])

# %%
from datetime import datetime
from datetime import timedelta
predict_data = dataset.tail(1)
time_predict_data = predict_data.index.values[0]
if (interval=="1m"):
    interval_time = 1
elif (interval == "5m"):
    interval_time = 5
elif (interval == "15m"):
    interval_time = 15
time_predict_data = pd.to_datetime(time_predict_data)+timedelta(minutes=interval_time)

predict_data = predict_data.values
predict_data = predict_data.astype('float32')

values = dataset.values
values = values.astype('float32')
# normalize features
scaler = MinMaxScaler(feature_range=(0, 1))
scaled = scaler.fit_transform(values)
scaled_predict = scaler.transform(predict_data)
# frame as supervised learning
reframed = series_to_supervised(scaled, 1, 1)
# drop columns we don't want to predict
# reframed.drop(reframed.columns[[9,10,11,12,13,14,15]], axis=1, inplace=True)
reframed = reframed.iloc[:, 0:(len(dataset.columns)+1)]
scaled_predict = DataFrame(scaled_predict)
scaled_predict.columns = reframed.columns[range(0, (len(dataset.columns)))]

# split into train and test sets
values = reframed.values
n_train_hours = 871
train = values[:n_train_hours, :]
test = values[n_train_hours:, :]
# split into input and outputs
train_X, train_y = train[:, :-1], train[:, -1]
test_X, test_y = test[:, :-1], test[:, -1]
# predict_test_X, predict_test_y = predict_test[:, :-1], predict_test[:, -1]

# reshape input to be 3D [samples, timesteps, features]
train_X = train_X.reshape((train_X.shape[0], 1, train_X.shape[1]))
test_X = test_X.reshape((test_X.shape[0], 1, test_X.shape[1]))
# predict_test_X = predict_test_X.reshape((predict_test_X.shape[0], 1, predict_test_X.shape[1]))
values_predict = scaled_predict.values
predict_data_x = values_predict.reshape((values_predict.shape[0], 1, values_predict.shape[1]))
print(train_X.shape, train_y.shape, test_X.shape, test_y.shape)
print(predict_data_x.shape)

# %%
# design network
from keras.layers import GRU
model = Sequential()
model.add(GRU(25, input_shape=(train_X.shape[1], train_X.shape[2])))
model.add(Dense(1))
model.compile(loss='mae', optimizer='adam')
history = model.fit(train_X, train_y, epochs=100, batch_size=32, validation_data=(test_X, test_y), verbose=1,
shuffle=False)

yhat_predict = model.predict(predict_data_x)

```

```

# predict_data_x
predict_data_x_new= predict_data_x.reshape((predict_data_x.shape[0], predict_data_x.shape[2]))

inv_yhat_predict_new = concatenate((yhat_predict, predict_data_x_new[:, 1:]), axis=1)
inv_yhat_predict_new = scaler.inverse_transform(inv_yhat_predict_new)
inv_yhat_predict_new = inv_yhat_predict_new[:,0].round(2)
# inv_yhat_predict_new = round(inv_yhat_predict_new,2)
print(time_predict_data)
print("%.2F" %inv_yhat_predict_new[0])

row_contents = [time_predict_data,inv_yhat_predict_new[0]]

from csv import writer
def append_list_as_row(file_name, list_of_elem):
    # Open file in append mode
    with open(file_name, 'a+', newline='') as write_obj:
        # Create a writer object from csv module
        csv_writer = writer(write_obj)
        # Add contents of list as last row in the csv file
        csv_writer.writerow(list_of_elem)

# Append a list as new line to an old csv file
append_list_as_row('result/btc-15m.csv', row_contents)

from deta import Deta

# 2) initialize with a project key
deta = Deta("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVFYX")

# 3) create and use as many DBs as you want!
post = deta.Base("btc15m_gru_db")

post.insert({
    "close_predict": str(inv_yhat_predict_new[0]),
    "time": str(time_predict_data)
})

```

## Lampiran 3 *Source code* scheduler model realtime

### Scheduler-1m.py

```

import time
import os

from datetime import datetime
import time
# starttime = datetime.now()
# print(starttime)
while True:
    os.system("python bnb-1m.py")
    starttime = datetime.now()
    time.sleep(63.0 - starttime.second)

```

### Scheduler-5m.py

```

import time
import os

from datetime import datetime
import time
# starttime = datetime.now()
# print(starttime)
while True:
    os.system("python bnb-5m.py")
    starttime = datetime.now()
    time.sleep((5 - starttime.minute%5)*60)
    time.sleep(3)

```

## Scheduler-15m.py

```
import time
import os

from datetime import datetime
import time
# starttime = datetime.now()
# print(starttime)
while True:
    os.system("python bnb-15m.py")
    starttime = datetime.now()
    time.sleep((15 - starttime.minute%15)*60)
    time.sleep(3)
```

## Lampiran 4 *Source code* evaluasi model

### tester.py

```
# %%
import pandas as pd

# %%
df_1m_day1 = pd.read_csv('bnb/gru/1m/bnb-day1.csv')
df_1m_day2 = pd.read_csv('bnb/gru/1m/bnb-day2.csv')
df_1m_day3 = pd.read_csv('bnb/gru/1m/bnb-day3.csv')
df_1m_day4 = pd.read_csv('bnb/gru/1m/bnb-day4.csv')
df_1m_day5 = pd.read_csv('bnb/gru/1m/bnb-day5.csv')
df_1m_day6 = pd.read_csv('bnb/gru/1m/bnb-day6.csv')
df_1m_day7 = pd.read_csv('bnb/gru/1m/bnb-day7.csv')

df_1m_day1['predict'].round(1)
df_1m_day2['predict'].round(1)
df_1m_day3['predict'].round(1)
df_1m_day4['predict'].round(1)
df_1m_day5['predict'].round(1)
df_1m_day6['predict'].round(1)
df_1m_day7['predict'].round(1)

# %%
#library
import pandas as pd
import time
import datetime
from datetime import timedelta
import math
import pytz
import pandas_ta as ta

# import urllib library
from urllib.request import urlopen
# import json
import json

def utcToUnix(timeData):
    timeData = datetime.datetime.timestamp(timeData)*1000
    timeData = math.trunc(timeData)
    # print(timeData)
    return timeData

def getData(timeStart,timeEnd,coin):
    # store the URL in url as
    # parameter for urlopen
    url =
    "https://api.binance.me/api/v3/klines?symbol="+str(coin)+"&interval=1m&limit=1000&startTime="+str(timeStart)+"&endTime="+str(timeEnd)
    # store the response of URL
    response = urlopen(url)

    # storing the JSON response
    # from url in data
    data_json = json.loads(response.read())

    # print the json response
    # print(data_json)
    tm,op,hi,lo,cl,v = [],[],[],[],[],[]

    for x in range(len(data_json)):
        # print(data_json[x][0])
        tm.append(data_json[x][0])
```

```

    op.append(data_json[x][1])
    hi.append(data_json[x][2])
    lo.append(data_json[x][3])
    cl.append(data_json[x][4])
    v.append(data_json[x][5])
    data = {"time": tm, "open":op,"high":hi,"low":lo,"close":cl,"volume":v}
    df = pd.DataFrame(data)
    df['time'] = df['time'].transform(lambda x: unix2utc(x))
    # df["open"] = pd.to_numeric(df["open"], downcast="float")
    # df["high"] = pd.to_numeric(df["high"], downcast="float")
    # df["low"] = pd.to_numeric(df["low"], downcast="float")
    df["close"] = pd.to_numeric(df["close"], downcast="float")
    # df["volume"] = pd.to_numeric(df["volume"], downcast="float")
    return df

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)
    return(utc_dt.strftime("%Y-%m-%d %H:%M:%S"))

def timeFrameOne(startTime,loop,timeFrame,interval,coin):
    dfall = pd.DataFrame()
    openTime = startTime
    for x in range(loop):
        openTime = openTime + datetime.timedelta(minutes = timeFrame)
        closeTime = openTime + datetime.timedelta(hours = interval)
        df = getData(utcToUnix(openTime),utcToUnix(closeTime),coin)
        # df = pd.concat(df)
        # print(len(df))
        dfall = dfall.append(df)
        # print(len(dfall))
        print(openTime)
        print(closeTime)
        openTime = closeTime
    print(len(dfall))
    json2data = json.loads(dfall.to_json(orient='records'))
    return json2data

day1 = datetime.datetime(2022, 7, 1, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day2 = datetime.datetime(2022, 7, 2, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day3 = datetime.datetime(2022, 7, 3, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day4 = datetime.datetime(2022, 7, 4, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day5 = datetime.datetime(2022, 7, 5, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day6 = datetime.datetime(2022, 7, 6, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)
day7 = datetime.datetime(2022, 7, 7, 16, 0, 0) + timedelta(hours=8)# format (yyyy, mm, dd, hh, mm, ss)

df_getday1 = timeFrameOne(startTime = day1, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday2 = timeFrameOne(startTime = day2, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday3 = timeFrameOne(startTime = day3, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday4 = timeFrameOne(startTime = day4, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday5 = timeFrameOne(startTime = day5, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday6 = timeFrameOne(startTime = day6, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")
df_getday7 = timeFrameOne(startTime = day7, loop = 2,interval = 16, timeFrame = 1, coin = "BNB")

data_day1 = pd.DataFrame(df_getday1)
data_day1 = data_day1.drop(columns=['open','high','low','volume'])
# data_day1['time'] = pd.to_datetime(data_day1['time'])

data_day2 = pd.DataFrame(df_getday2)
data_day2 = data_day2.drop(columns=['open','high','low','volume'])
# data_day2['time'] = pd.to_datetime(data_day2['time'])

data_day3 = pd.DataFrame(df_getday3)
data_day3 = data_day3.drop(columns=['open','high','low','volume'])
# data_day3['time'] = pd.to_datetime(data_day3['time'])

data_day4 = pd.DataFrame(df_getday4)
data_day4 = data_day4.drop(columns=['open','high','low','volume'])
# data_day4['time'] = pd.to_datetime(data_day4['time'])

data_day5 = pd.DataFrame(df_getday5)
data_day5 = data_day5.drop(columns=['open','high','low','volume'])
# data_day5['time'] = pd.to_datetime(data_day5['time'])

data_day6 = pd.DataFrame(df_getday6)
data_day6 = data_day6.drop(columns=['open','high','low','volume'])
# data_day5['time'] = pd.to_datetime(data_day5['time'])

data_day7 = pd.DataFrame(df_getday7)
data_day7 = data_day7.drop(columns=['open','high','low','volume'])
# data_day5['time'] = pd.to_datetime(data_day5['time'])

# %%
new_data_day1 = pd.merge(data_day1,df_1m_day1, on='time')
new_data_day1['predict'] = new_data_day1['predict'].transform(lambda x: round(x,1))
new_data_day1['close'] = new_data_day1['close'].transform(lambda x: round(x,1))
new_data_day1['time'] = pd.to_datetime(new_data_day1['time'])

new_data_day2 = pd.merge(data_day2,df_1m_day2, on='time')

```

```

new_data_day2['predict'] = new_data_day2['predict'].transform(lambda x: round(x,1))
new_data_day2['close'] = new_data_day2['close'].transform(lambda x: round(x,1))
new_data_day2['time'] = pd.to_datetime(new_data_day2['time'])

new_data_day3 = pd.merge(data_day3,df_1m_day3, on='time')
new_data_day3['predict'] = new_data_day3['predict'].transform(lambda x: round(x,1))
new_data_day3['close'] = new_data_day3['close'].transform(lambda x: round(x,1))
new_data_day3['time'] = pd.to_datetime(new_data_day3['time'])

new_data_day4 = pd.merge(data_day4,df_1m_day4, on='time')
new_data_day4['predict'] = new_data_day4['predict'].transform(lambda x: round(x,1))
new_data_day4['close'] = new_data_day4['close'].transform(lambda x: round(x,1))
new_data_day4['time'] = pd.to_datetime(new_data_day4['time'])

new_data_day5 = pd.merge(data_day5,df_1m_day5, on='time')
new_data_day5['predict'] = new_data_day5['predict'].transform(lambda x: round(x,1))
new_data_day5['close'] = new_data_day5['close'].transform(lambda x: round(x,1))
new_data_day5['time'] = pd.to_datetime(new_data_day5['time'])

new_data_day6 = pd.merge(data_day6,df_1m_day6, on='time')
new_data_day6['predict'] = new_data_day6['predict'].transform(lambda x: round(x,1))
new_data_day6['close'] = new_data_day6['close'].transform(lambda x: round(x,1))
new_data_day6['time'] = pd.to_datetime(new_data_day6['time'])

new_data_day7 = pd.merge(data_day7,df_1m_day7, on='time')
new_data_day7['predict'] = new_data_day7['predict'].transform(lambda x: round(x,1))
new_data_day7['close'] = new_data_day7['close'].transform(lambda x: round(x,1))
new_data_day7['time'] = pd.to_datetime(new_data_day7['time'])

frames = [new_data_day1,new_data_day2,new_data_day3,new_data_day4,new_data_day5,new_data_day6,new_data_day7]
all_data = pd.concat(frames)

# %%
all_data

# %%
import numpy as np
from sklearn.metrics import mean_squared_error
from math import sqrt

def mape(actual, pred):
    actual, pred = np.array(actual), np.array(pred)
    return np.mean(np.abs((actual - pred) / actual)) * 100

rmse_day1 = sqrt(mean_squared_error(new_data_day1['close'], new_data_day1['predict']))
mape_day1 = mape(new_data_day1['predict'], new_data_day1['close'])

rmse_day2 = sqrt(mean_squared_error(new_data_day2['close'], new_data_day2['predict']))
mape_day2 = mape(new_data_day2['close'], new_data_day2['predict'])

rmse_day3 = sqrt(mean_squared_error(new_data_day3['close'], new_data_day3['predict']))
mape_day3 = mape(new_data_day3['close'], new_data_day3['predict'])

rmse_day4 = sqrt(mean_squared_error(new_data_day4['close'], new_data_day4['predict']))
mape_day4 = mape(new_data_day4['close'], new_data_day4['predict'])

rmse_day5 = sqrt(mean_squared_error(new_data_day5['close'], new_data_day5['predict']))
mape_day5 = mape(new_data_day5['close'], new_data_day5['predict'])

rmse_day6 = sqrt(mean_squared_error(new_data_day6['close'], new_data_day6['predict']))
mape_day6 = mape(new_data_day6['close'], new_data_day6['predict'])

rmse_day7 = sqrt(mean_squared_error(new_data_day7['close'], new_data_day7['predict']))
mape_day7 = mape(new_data_day7['close'], new_data_day7['predict'])

rmse_all = sqrt(mean_squared_error(all_data['close'], all_data['predict']))
mape_all = mape(all_data['close'], all_data['predict'])

# %%
import matplotlib.pyplot as plt

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 1\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day1 ,mape_day1))
plt.plot(new_data_day1['time'],new_data_day1['close'],label = 'actual')
plt.plot(new_data_day1['time'],new_data_day1['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 2\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day2 ,mape_day2))
plt.plot(new_data_day2['time'],new_data_day2['close'],label = 'actual')
plt.plot(new_data_day2['time'],new_data_day2['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))

```

```

# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 3\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day3 ,mape_day3))
plt.plot(new_data_day3['time'],new_data_day3['close'],label = 'actual')
plt.plot(new_data_day3['time'],new_data_day3['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 4\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day4 ,mape_day4))
plt.plot(new_data_day4['time'],new_data_day4['close'],label = 'actual')
plt.plot(new_data_day4['time'],new_data_day4['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 5\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day5 ,mape_day5))
plt.plot(new_data_day5['time'],new_data_day5['close'],label = 'actual')
plt.plot(new_data_day5['time'],new_data_day5['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 6\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day6 ,mape_day6))
plt.plot(new_data_day6['time'],new_data_day6['close'],label = 'actual')
plt.plot(new_data_day6['time'],new_data_day6['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m day 7\nRMSE : %.3f\nMAPE : %.3f" %(rmse_day7 ,mape_day7))
plt.plot(new_data_day7['time'],new_data_day7['close'],label = 'actual')
plt.plot(new_data_day7['time'],new_data_day7['predict'],label = 'predict')
plt.legend()

plt.figure(figsize=(10, 5))
# plt.figure()
plt.title("[GRU] BNB timeframe 1m 7 Days\nRMSE : %.3f\nMAPE : %.3f" %(rmse_all ,mape_all))
plt.plot(all_data['time'],all_data['close'],label = 'actual')
plt.plot(all_data['time'],all_data['predict'],label = 'predict')
plt.legend()

# %%
all_data.to_csv("csv/gru_bnb_1m.csv")

# %%

```

## Lampiran 5 Sourcecode Microservice

```

from doctest import Example
from fastapi import FastAPI,Body
from fastapi.middleware.cors import CORSMiddleware
from data import Data
import pandas as pd
from typing import Optional
import json
import math
import datetime
from datetime import timedelta
from urllib.request import urlopen
import pandas as pd
import pytz
from cryptomodule import unix2utc,getDataCrypto,utcToUnix

import crypto_module as cm

app = FastAPI()
origins = ["*"]

app.add_middleware(
    CORSMiddleware,
    allow_origins=origins,
    allow_credentials=True,
    allow_methods=["*"],
    allow_headers=["*"],
)

def utcToUnix(timeData):
    timeData = datetime.datetime.timestamp(timeData)*1000
    timeData = math.trunc(timeData)
    return timeData

data = Data("c0fws2vv_pM7b7NMZEYyaCpSJBnv3W1tsmpGSVFYX")

def unix2utc(timestamp):
    your_dt = datetime.datetime.fromtimestamp(int(timestamp)/1000) # using the local timezone
    utc_dt= your_dt.astimezone(pytz.UTC)

```



```

return(utc_dt.strftime("%Y-%m-%d %H:%M:%S")) # 2018-04-07 20:48:08, YMMV

@app.get("/")
async def root():
    return {"test": "OK", "docs": "https://fanfan-api.deta.dev/docs"}

@app.get("/bnb")
async def bnb(timeframe:str,limit:Optional[str] = None):

    tf = "bnb"+ timeframe + "_db"
    tf2 = "bnb"+ timeframe + "_gru_db"
    bnb = deta.Base(tf)
    bnb_gru = deta.Base(tf2)

    if(timeframe == '1m'):
        datetimenow = datetime.datetime.now() - timedelta(hours=2)
    else:
        datetimenow = datetime.datetime.now() - timedelta(hours=20)

    datetimenow1 = datetime.datetime.now()+timedelta(minutes=20)

    if(timeframe == '1m'):
        res = bnb.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
    else:
        res = bnb.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)

    all_item = res.items
    all_item_gru = res_gru.items
    # while res.last:
    #     res = bnb.fetch(last=res.last)
    #     all_item+=res.items

    # while res_gru.last:
    #     res_gru = bnb_gru.fetch(last=res_gru.last)
    #     all_item_gru+=res_gru.items

    data = pd.DataFrame(all_item)
    data_gru = pd.DataFrame(all_item_gru)

    data['time'] = data['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))
    data_gru['time'] = data_gru['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))

    data.rename(columns ={'close_predict':'close_predict_lstm'}, inplace = True)
    data_gru.rename(columns ={'close_predict':'close_predict_gru'}, inplace = True)

    data =data[['time','close_predict_lstm']]
    data_gru =data_gru[['time','close_predict_gru']]

    data = data.sort_values(by="time")
    data['time'] = data['time'].transform(lambda x: str(x))
    data = data.reset_index()
    # data = data.set_index('time')
    data = data.iloc[:,1:]

    data_gru = data_gru.sort_values(by="time")
    data_gru['time'] = data_gru['time'].transform(lambda x: str(x))
    data_gru = data_gru.reset_index()
    # data = data.set_index('time')
    data_gru = data_gru.iloc[:,1:]

    if(limit is None):
        data = data.tail(5)
        data_gru = data_gru.tail(5)
    elif(int(limit) < 1000):
        data = data.tail(int(limit))
        data_gru = data_gru.tail(int(limit))

    data = pd.merge(data,data_gru, on='time',how='outer')
    currency = "BNBBUSD"
    interval = "1m"
    len_data = "1000"
    start_time = data.head(1)
    start_time = start_time['time'].values[0]
    print(start_time)
    start_time_dt = datetime.datetime.strptime(start_time, "%Y-%m-%d %H:%M:%S") + timedelta(hours=8)
    start_time_dt = str(utcToUnix(start_time_dt))
    df_scrap = cm.getDataCrypto(currency,interval,len_data,start_time_dt,None)
    df_scrap = df_scrap[['time','close']]
    df_scrap = pd.DataFrame(df_scrap)
    df_scrap['time'] = df_scrap['time'].transform(lambda x: unix2utc(x))
    df_scrap['close'] = df_scrap['close'].transform(lambda x: round(x,2))
    new_data = pd.merge(data,df_scrap, on='time',how='left')
    new_data = new_data.fillna('TBD')
    new_data['close'] = new_data['close'].transform(lambda x: str(x))
    json2data = json.loads(new_data.to_json(orient='records'))

```

```

json2scrap = json.loads(df_scrap.to_json(orient='records'))
# json2data = json.loads(data.to_json(orient='records'))
return { "data": json2data}

@app.get("/eth")
async def eth(timeframe:str,limit:Optional[str] = None):
    tf = "eth"+ timeframe + "_db"
    tf2 = "eth"+ timeframe + "_gru_db"
    bnb = deta.Base(tf)
    bnb_gru = deta.Base(tf2)

    if(timeframe == '1m'):
        datetimenow = datetime.datetime.now() - timedelta(hours=2)
    else:
        datetimenow = datetime.datetime.now() - timedelta(hours=20)

    datetimenow1 = datetime.datetime.now()+timedelta(minutes=20)

    if(timeframe == '1m'):
        res = bnb.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
    else:
        res = bnb.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)

    all_item = res.items
    all_item_gru = res_gru.items
    # while res.last:
    #     res = bnb.fetch(last=res.last)
    #     all_item+=res.items

    # while res_gru.last:
    #     res_gru = bnb_gru.fetch(last=res_gru.last)
    #     all_item_gru+=res_gru.items

    data = pd.DataFrame(all_item)
    data_gru = pd.DataFrame(all_item_gru)

    data['time'] = data['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))
    data_gru['time'] = data_gru['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))

    data.rename(columns ={'close_predict':'close_predict_lstm'}, inplace = True)
    data_gru.rename(columns ={'close_predict':'close_predict_gru'}, inplace = True)

    data =data[['time','close_predict_lstm']]
    data_gru =data_gru[['time','close_predict_gru']]

    data = data.sort_values(by="time")
    data['time'] = data['time'].transform(lambda x: str(x))
    data = data.reset_index()
    # data = data.set_index('time')
    data = data.iloc[:,1:]

    data_gru = data_gru.sort_values(by="time")
    data_gru['time'] = data_gru['time'].transform(lambda x: str(x))
    data_gru = data_gru.reset_index()
    # data = data.set_index('time')
    data_gru = data_gru.iloc[:,1:]

    if(limit is None):
        data = data.tail(5)
        data_gru = data_gru.tail(5)
    elif(int(limit) < 1000):
        data = data.tail(int(limit))
        data_gru = data_gru.tail(int(limit))

    data = pd.merge(data,data_gru, on='time',how='outer')
    currency = "ETHBUSD"
    interval = "1m"
    len_data = "1000"
    start_time = data.head(1)
    start_time = start_time['time'].values[0]
    print(start_time)
    start_time_dt = datetime.datetime.strptime(start_time, "%Y-%m-%d %H:%M:%S") + timedelta(hours=8)
    start_time_dt = str(utcToUnix(start_time_dt))
    df_scrap = cm.getDataCrypto(currency,interval,len_data,start_time_dt,None)
    df_scrap = df_scrap[['time','close']]
    df_scrap = pd.DataFrame(df_scrap)
    df_scrap['time'] = df_scrap['time'].transform(lambda x: unix2utc(x))
    df_scrap['close'] = df_scrap['close'].transform(lambda x: round(x,2))
    new_data = pd.merge(data,df_scrap, on='time',how='left')
    new_data = new_data.fillna('TBD')
    new_data['close'] = new_data['close'].transform(lambda x: str(x))
    json2data = json.loads(new_data.to_json(orient='records'))
    json2scrap = json.loads(df_scrap.to_json(orient='records'))
    # json2data = json.loads(data.to_json(orient='records'))
    return { "data": json2data}

```

```

@app.get("/btc")
async def btc(timeframe:str,limit:Optional[str] = None):
    tf = "btc"+ timeframe + "_db"
    tf2 = "btc"+ timeframe + "_gru_db"
    bnb = deta.Base(tf)
    bnb_gru = deta.Base(tf2)

    if(timeframe == '1m'):
        datetimenow = datetime.datetime.now() - timedelta(hours=2)
    else:
        datetimenow = datetime.datetime.now() - timedelta(hours=20)

    datetimenow1 = datetime.datetime.now()+timedelta(minutes=20)

    if(timeframe == '1m'):
        res = bnb.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"key?r": [str(utcToUnix(datetimenow)),str(utcToUnix(datetimenow1))]}],limit=1000)
    else:
        res = bnb.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)
        res_gru = bnb_gru.fetch(query=[{"time?r": [str(datetimenow),str(datetimenow1)]}],limit=1000)

    all_item = res.items
    all_item_gru = res_gru.items
    # while res.last:
    #     res = bnb.fetch(last=res.last)
    #     all_item+=res.items

    # while res_gru.last:
    #     res_gru = bnb_gru.fetch(last=res_gru.last)
    #     all_item_gru+=res_gru.items

    data = pd.DataFrame(all_item)
    data_gru = pd.DataFrame(all_item_gru)

    data['time'] = data['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))
    data_gru['time'] = data_gru['time'].transform(lambda x: datetime.datetime.strptime(x, '%Y-%m-%d %H:%M:%S'))

    data.rename(columns ={'close_predict':'close_predict_lstm'}, inplace = True)
    data_gru.rename(columns ={'close_predict':'close_predict_gru'}, inplace = True)

    data =data[['time', 'close_predict_lstm']]
    data_gru =data_gru[['time', 'close_predict_gru']]

    data = data.sort_values(by="time")
    data['time'] = data['time'].transform(lambda x: str(x))
    data = data.reset_index()
    # data = data.set_index('time')
    data = data.iloc[:,1:]

    data_gru = data_gru.sort_values(by="time")
    data_gru['time'] = data_gru['time'].transform(lambda x: str(x))
    data_gru = data_gru.reset_index()
    # data = data.set_index('time')
    data_gru = data_gru.iloc[:,1:]

    if(limit is None):
        data = data.tail(5)
        data_gru = data_gru.tail(5)
    elif(int(limit) < 1000):
        data = data.tail(int(limit))
        data_gru = data_gru.tail(int(limit))

    data = pd.merge(data,data_gru, on='time',how='outer')
    currency = "BTCUSD"
    interval = "1m"
    len_data = "1000"
    start_time = data.head(1)
    start_time = start_time['time'].values[0]
    print(start_time)
    start_time_dt = datetime.datetime.strptime(start_time, "%Y-%m-%d %H:%M:%S") + timedelta(hours=8)
    start_time_dt = str(utcToUnix(start_time_dt))
    df_scrap = cm.getDataCrypto(currency,interval,len_data,start_time_dt,None)
    df_scrap = df_scrap[['time','close']]
    df_scrap = pd.DataFrame(df_scrap)
    df_scrap['time'] = df_scrap['time'].transform(lambda x: unix2utc(x))
    df_scrap['close'] = df_scrap['close'].transform(lambda x: round(x,2))
    new_data = pd.merge(data,df_scrap, on='time',how='left')
    new_data = new_data.fillna('TBD')
    new_data['close'] = new_data['close'].transform(lambda x: str(x))
    json2data = json.loads(new_data.to_json(orient='records'))
    json2scrapp = json.loads(df_scrap.to_json(orient='records'))
    # json2data = json.loads(data.to_json(orient='records'))
    return { "data": json2data}
@app.get("/testing")

```

```

async def testing():
    currency = "BTCBUSD"
    interval = "1m"
    len_data = "100"
    start_time = "2022-07-20 00:41:00"
    print(start_time)
    start_time_dt = datetime.datetime.strptime(start_time, "%Y-%m-%d %H:%M:%S") + timedelta(hours=8)
    start_time_dt = str(utcToUnix(start_time_dt))
    # df_scrap = getDataCrypto(currency,interval,len_data,start_time_dt)
    df_scrap = cm.getDataCrypto(currency,interval,len_data,start_time_dt,None)
    df_scrap = df_scrap[['time','close']]
    json2scraper = json.loads(df_scrap.to_json(orient='records'))
    return {"scraper":json2scraper}

```

## Lampiran 6 Sourcecode Dashboard BTC

```

<!doctype html>
<html lang="en">

<head>
<meta charset="utf-8">
<meta name="viewport" content="width=device-width, initial-scale=1">
<meta name="description" content="">
<meta name="author" content="Mark Otto, Jacob Thornton, and Bootstrap contributors">
<meta name="generator" content="Hugo 0.84.0">
<title>Fanfan Crypto</title>
<script src="https://code.jquery.com/jquery-3.2.1.min.js"></script>
<link rel="canonical" href="https://getbootstrap.com/docs/5.0/examples/product/">
<!-- Bootstrap core CSS -->
<link href="assets/dist/css/bootstrap.min.css" rel="stylesheet">
<style>
    .bd-placeholder-img {
        font-size: 1.125rem;
        text-align: middle;
        -webkit-user-select: none;
        -moz-user-select: none;
        user-select: none;
    }

    @media (min-width: 768px) {
        .bd-placeholder-img-lg {
            font-size: 3.5rem;
        }
    }
</style>

<!-- Custom styles for this template -->
<link href="product.css" rel="stylesheet">
</head>
<body>
<header class="site-header sticky-top py-1">
<nav class="container d-flex flex-column flex-md-row justify-content-between">
<a class="py-2" href="#" aria-label="Product">
<svg xmlns="http://www.w3.org/2000/svg" width="24" height="24" fill="none" stroke="currentColor"
stroke-linecap="round" stroke-linejoin="round" stroke-width="2" class="d-block mx-auto" role="img"
viewBox="0 0 24 24">
<title>Product</title>
<circle cx="12" cy="12" r="10" />
<path
d="M14.31 815.74 9.94M9.69 8h11.48M7.38 1215.74-9.94M9.69 16L3.95 6.06M14.31 16H2.83m13.79-41-5.74 9.94"
/>
</svg>
</a>
<a class="py-2 d-none d-md-inline-block" href="index.html">BTC/BUSD</a>
<a class="py-2 d-none d-md-inline-block" href="eth.html">ETH/BUSD</a>
<a class="py-2 d-none d-md-inline-block" href="bnb.html">BNB/BUSD</a>
</nav>
</header>

<main>
<div class="position-relative overflow-hidden p-3 p-md-5 m-md-3 text-center bg-light">
<div class="col-md-5 p-lg-5 mx-auto my-5">
<h1 class="display-4 fw-normal">fanfanmyid</h1>
<p class="lead fw-normal">Prediksi Nilai Tukar Cryptocurrency menggunakan LSTM dan GRU</p>
</div>
</div>
<div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 py-3" id="btc_banner">
<!-- <p class="lead btc_time" id="time_btc">19.20.</p>
<p class="lead" id="predict_btc">And an even wittier subheading.</p> -->
</div>

```

```

</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
  <div class="my-3 p-3" id="eth_banner">
    <!-- <h5 class="display-8">ETH/BUSD</h5>
    <p class="lead">And an even wittier subheading.</p> -->
  </div>
</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
  <div class="my-3 p-3" id="bnb_banner">
    <!-- <h5 class="display-8">BNB/BUSD</h5>
    <p class="lead">And an even wittier subheading.</p> -->
  </div>
</div>
</div>
<div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
  <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
    <div class="my-3 py-3">
      <H4>BTC/BUSD 1m</H4>
      <table class="table">
        <thead>
          <tr>
            <th scope="col">Time</th>
            <th scope="col">Predict LSTM</th>
            <th scope="col">Predict GRU</th>
            <th scope="col">Actual</th>
          </tr>
        </thead>
        <tbody id="btcdata">
        </tbody>
      </table>
    </div>

  </div>

  <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
    <div class="my-3 p-3">
      <H4>BTC/BUSD 5m</H4>
      <table class="table">
        <thead>
          <tr>
            <th scope="col">Time</th>
            <th scope="col">Predict LSTM</th>
            <th scope="col">Predict GRU</th>
            <th scope="col">Actual</th>
          </tr>
        </thead>
        <tbody id="ethdata">
        </tbody>
      </table>
    </div>

  </div>

  <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
    <div class="my-3 p-3">
      <H4>BTC/BUSD 15m</H4>
      <table class="table">
        <thead>
          <tr>
            <th scope="col">Time</th>
            <th scope="col">Predict LSTM</th>
            <th scope="col">Predict GRU</th>
            <th scope="col">Actual</th>
          </tr>
        </thead>
        <tbody id="bnldata">
        </tbody>
      </table>
    </div>

  </div>
</div>
</div>
<script>
function onBtcGet() {
  const url = "https://fanfan-api deta.dev/btc?timeframe=1m";
  var headers = {}

  fetch(url)
    .then((response) => {
      if (!response.ok) {
        throw new Error(response.error)
      }
      return response.json();
    })
    .then(data => {
      var jsonData = data.data
      var temp = ""
      var banner = ""

```

```

for (let x in jsonData) {
  temp += "<tr>";
  date = new Date(Date.parse(jsonData[x]['time']+" UTC"));
  temp += "<td>" + date.toLocaleString() + "</td>";
  temp += "<td>" + jsonData[x]['close_predict_lstm'] + "</td>";
  temp += "<td>" + jsonData[x]['close_predict_gru'] + "</td>";
  temp += "<td>" + jsonData[x]['close'] + "</td>";
  if (x == 4){
    console.log("last data")
    banner+="

##### 


```

```

temp += "<td>" + jsonData[x]['close'] + "</td>";
if (x == 4){
  banner+="

##### 


```

## Lampiran 7 Sourcecode Dashboard ETH

```

<!doctype html>
<html lang="en">

<head>
<meta charset="utf-8">
<meta name="viewport" content="width=device-width, initial-scale=1">
<meta name="description" content="">
<meta name="author" content="Mark Otto, Jacob Thornton, and Bootstrap contributors">
<meta name="generator" content="Hugo 0.84.0">
<title>Fanfan Crypto</title>
<script src="https://code.jquery.com/jquery-3.2.1.min.js"></script>
<link rel="canonical" href="https://getbootstrap.com/docs/5.0/examples/product/">
<!-- Bootstrap core CSS -->
<link href="assets/dist/css/bootstrap.min.css" rel="stylesheet">

<style>
.bd-placeholder-img {
  font-size: 1.125rem;
  text-anchor: middle;
  -webkit-user-select: none;
  -moz-user-select: none;
  user-select: none;
}

@media (min-width: 768px) {
  .bd-placeholder-img-lg {
    font-size: 3.5rem;
  }
}
</style>

<!-- Custom styles for this template -->
<link href="product.css" rel="stylesheet">
</head>

<body>

<header class="site-header sticky-top py-1">
<nav class="container d-flex flex-column flex-md-row justify-content-between">
  <a class="py-2" href="#" aria-label="Product">
    <svg xmlns="http://www.w3.org/2000/svg" width="24" height="24" fill="none" stroke="currentColor"

```

```

        stroke-linecap="round" stroke-linejoin="round" stroke-width="2" class="d-block mx-auto" role="img"
        viewBox="0 0 24 24">
        <title>Product</title>
        <circle cx="12" cy="12" r="10" />
        <path
        d="M14.31 815.74 9.94M9.69 8h11.48M7.38 1215.74-9.94M9.69 16L14.95 6.06M14.31 16H2.83m13.79-41-5.74 9.94"
        />
    </svg>
    </a>
    <a class="py-2 d-none d-md-inline-block" href="index.html">BTC/BUSD</a>
    <a class="py-2 d-none d-md-inline-block" href="eth.html">ETH/BUSD</a>
    <a class="py-2 d-none d-md-inline-block" href="bnb.html">BNB/BUSD</a>
</nav>
</header>
<main>
<div class="position-relative overflow-hidden p-3 p-md-5 m-md-3 text-center bg-light">
<div class="col-md-5 p-lg-5 mx-auto my-5">
<h1 class="display-4 fw-normal">fanfanmyid</h1>
<p class="lead fw-normal">Prediksi Nilai Tukar Cryptocurrency menggunakan LSTM dan GRU</p>
</div>
</div>
<div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 py-3" id="btc_banner">
<!-- <p class="lead btc_time" id="time_btc">19.20.</p>
<p class="lead" id="predict_btc">And an even wittier subheading.</p -->
</div>
</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 p-3" id="eth_banner">
<!-- <h5 class="display-8">ETH/BUSD</h5>
<p class="lead">And an even wittier subheading.</p -->
</div>
</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 p-3" id="bnb_banner">
<!-- <h5 class="display-8">BNB/BUSD</h5>
<p class="lead">And an even wittier subheading.</p -->
</div>
</div>
</div>
<div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 py-3">
<H4>ETH/BUSD 1m</H4>
<table class="table">
<thead>
<tr>
<th scope="col">Time</th>
<th scope="col">Predict LSTM</th>
<th scope="col">Predict GRU</th>
<th scope="col">Actual</th>
</tr>
</thead>
<tbody id="btcdata">
</tbody>
</table>
</div>
</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 p-3">
<H4>ETH/BUSD 5m</H4>
<table class="table">
<thead>
<tr>
<th scope="col">Time</th>
<th scope="col">Predict LSTM</th>
<th scope="col">Predict GRU</th>
<th scope="col">Actual</th>
</tr>
</thead>
<tbody id="ethdata">
</tbody>
</table>
</div>
</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
<div class="my-3 p-3">
<H4>ETH/BUSD 15m</H4>
<table class="table">
<thead>
<tr>
<th scope="col">Time</th>
<th scope="col">Predict LSTM</th>

```



```

        <th scope="col">Predict GRU</th>
        <th scope="col">Actual</th>
    </tr>
</thead>
<tbody id="bnbdata">
</tbody>
</table>
</div>
</div>
</div>
<script>
function onBtcGet() {
const url = "https://fanfan-api.deta.dev/eth?timeframe=1m";
var headers = {}
fetch(url)
    .then((response) => {
        if (!response.ok) {
            throw new Error(response.error)
        }
        return response.json();
    })
    .then(data => {
        var jsonData = data.data
        var temp = ""
        var banner = ""
        for (let x in jsonData) {
            temp += "<tr>";
            date = new Date(Date.parse(jsonData[x]['time']+" UTC"))
            temp += "<td>" + date.toLocaleString() + "</td>";
            temp += "<td>" + jsonData[x]['close_predict_lstm'] + "</td>";
            temp += "<td>" + jsonData[x]['close_predict_gru'] + "</td>";
            temp += "<td>" + jsonData[x]['close'] + "</td>";
            if (x == 4){
                console.log("last data")
                banner+="

##### 


```

```

}
function onBnbGet() {
  const url = "https://fanfan-api.deta.dev/eth?timeframe=15m";
  var headers = {}
  fetch(url)
    .then((response) => {
      if (!response.ok) {
        throw new Error(response.error)
      }
      return response.json();
    })
    .then(data => {
      var jsonData = data.data
      var temp = ""
      var banner = ""
      console.log(jsonData.length)
      for (let x in jsonData) {
        temp += "<tr>";
        date = new Date(Date.parse(jsonData[x]['time']+" UTC"))
        temp += "<td>" + date.toLocaleString() + "</td>";
        temp += "<td>" + jsonData[x]['close_predict_lstm'] + "</td>";
        temp += "<td>" + jsonData[x]['close_predict_gru'] + "</td>";
        temp += "<td>" + jsonData[x]['close'] + "</td>";
        if (x == 4){
          banner+="

##### 


```

## Lampiran 8 Sourcecode Dashboard BNB

```

<!doctype html>
<html lang="en">
<head>
  <meta charset="utf-8">
  <meta name="viewport" content="width=device-width, initial-scale=1">
  <meta name="description" content="">
  <meta name="author" content="Mark Otto, Jacob Thornton, and Bootstrap contributors">
  <meta name="generator" content="Hugo 0.84.0">
  <title>Fanfan Crypto</title>
  <script src="https://code.jquery.com/jquery-3.2.1.min.js"></script>
  <link rel="canonical" href="https://getbootstrap.com/docs/5.0/examples/product/">
  <link href="assets/dist/css/bootstrap.min.css" rel="stylesheet">
  <style>
    .bd-placeholder-img {
      font-size: 1.125rem;
      text-anchor: middle;
      -webkit-user-select: none;
      -moz-user-select: none;
      user-select: none;
    }
  </style>

```

```

@media (min-width: 768px) {
  .bd-placeholder-img-lg {
    font-size: 3.5rem;
  }
}
</style>
<!-- Custom styles for this template -->
<link href="product.css" rel="stylesheet">
</head>
<body>
  <header class="site-header sticky-top py-1">
    <nav class="container d-flex flex-column flex-md-row justify-content-between">
      <a class="py-2" href="#" aria-label="Product">
        <svg xmlns="http://www.w3.org/2000/svg" width="24" height="24" fill="none" stroke="currentColor"
          stroke-linecap="round" stroke-linejoin="round" stroke-width="2" class="d-block mx-auto" role="img"
          viewBox="0 0 24 24">
          <title>Product</title>
          <circle cx="12" cy="12" r="10" />
          <path
            d="M14.31 815.74 9.94M9.69 8h11.48M7.38 1215.74-9.94M9.69 16L3.95 6.06M14.31 16H2.83m13.79-41-5.74 9.94"
          />
        </svg>
      </a>
      <a class="py-2 d-none d-md-inline-block" href="index.html">BTC/BUSD</a>
      <a class="py-2 d-none d-md-inline-block" href="eth.html">ETH/BUSD</a>
      <a class="py-2 d-none d-md-inline-block" href="bnb.html">BNB/BUSD</a>
    </nav>
  </header>
  <main>
    <div class="position-relative overflow-hidden p-3 p-md-5 m-md-3 text-center bg-light">
      <div class="col-md-5 p-lg-5 mx-auto my-5">
        <h1 class="display-4 fw-normal">fanfanmyid</h1>
        <p class="lead fw-normal">Prediksi Nilai Tukar Cryptocurrency menggunakan LSTM dan GRU</p>
      </div>
      <div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
        <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
          <div class="my-3 py-3" id="btc_banner">
            <!-- <p class="lead btc_time" id="time_btc">19.20.</p>
            <p class="lead" id="predict_btc">And an even wittier subheading.</p> -->
          </div>
        </div>
        <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
          <div class="my-3 p-3" id="eth_banner">
            <!-- <h5 class="display-8">ETH/BUSD</h5>
            <p class="lead">And an even wittier subheading.</p> -->
          </div>
        </div>
        <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
          <div class="my-3 p-3" id="bnb_banner">
            <!-- <h5 class="display-8">BNB/BUSD</h5>
            <p class="lead">And an even wittier subheading.</p> -->
          </div>
        </div>
      </div>
      <div class="d-md-flex flex-md-equal w-100 my-md-3 ps-md-3">
        <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
          <div class="my-3 py-3">
            <H4>BNB/BUSD 1m</H4>
            <table class="table">
              <thead>
                <tr>
                  <th scope="col">Time</th>
                  <th scope="col">Predict LSTM</th>
                  <th scope="col">Predict GRU</th>
                  <th scope="col">Actual</th>
                </tr>
              </thead>
              <tbody id="btcdata">
                </tbody>
            </table>
          </div>
        </div>
        <div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
          <div class="my-3 p-3">
            <H4>BNB/BUSD 5m</H4>
            <table class="table">
              <thead>
                <tr>
                  <th scope="col">Time</th>
                  <th scope="col">Predict LSTM</th>
                  <th scope="col">Predict GRU</th>
                  <th scope="col">Actual</th>
                </tr>
              </thead>
              <tbody id="ethdata">
                </tbody>
            </table>
          </div>
        </div>
      </div>
    </div>
  </main>

```

```

</div>
<div class="bg-light me-md-3 pt-3 px-3 pt-md-5 px-md-5 text-center overflow-hidden">
  <div class="my-3 p-3">
    <h4>BNB/BUSD 15m</h4>
    <table class="table">
      <thead>
        <tr>
          <th scope="col">Time</th>
          <th scope="col">Predict LSTM</th>
          <th scope="col">Predict GRU</th>
          <th scope="col">Actual</th>
        </tr>
      </thead>
      <tbody id="bnbdata">
      </tbody>
    </table>
  </div>
</div>
</div>
<script>
function onBtcGet() {
  const url = "https://fanfan-api.deta.dev/bnb?timeframe=1m";
  var headers = {}
  fetch(url)
    .then((response) => {
      if (!response.ok) {
        throw new Error(response.error)
      }
      return response.json();
    })
    .then(data => {
      var jsonData = data.data
      var temp = ""
      var banner = ""
      for (let x in jsonData) {
        temp += "<tr>";
        date = new Date(Date.parse(jsonData[x]['time']+" UTC"))
        temp += "<td>" + date.toLocaleString() + "</td>";
        temp += "<td>" + jsonData[x]['close_predict_lstm'] + "</td>";
        temp += "<td>" + jsonData[x]['close_predict_gru'] + "</td>";
        temp += "<td>" + jsonData[x]['close'] + "</td>";
        if (x == 4){
          console.log("last data")
          banner+="

##### 


```

```

        banner += "<p class='lead'>" + "Predict GRU :." + "</p>"
        banner += "<p class='lead'>" + jsonData[x]['close_predict_gru'] + "</p>"
        document.getElementById('eth_banner').innerHTML = banner
    }
}
document.getElementById('ethdata').innerHTML = temp;
})
.catch(function (error) {
    console.log(error)
});
}
function onBnbGet() {
    const url = "https://fanfan-api.deta.dev/bnb?timeframe=15m";
    var headers = {}
    fetch(url)
        .then((response) => {
            if (!response.ok) {
                throw new Error(response.error)
            }
            return response.json();
        })
        .then(data => {
            var jsonData = data.data
            var temp = ""
            var banner = ""
            console.log(jsonData.length)
            for (let x in jsonData) {
                temp += "<tr>";
                date = new Date(Date.parse(jsonData[x]['time']+" UTC"))
                temp += "<td>" + date.toLocaleString() + "</td>";
                temp += "<td>" + jsonData[x]['close_predict_lstm'] + "</td>";
                temp += "<td>" + jsonData[x]['close_predict_gru'] + "</td>";
                temp += "<td>" + jsonData[x]['close'] + "</td>";
                if (x == 4){
                    banner+="

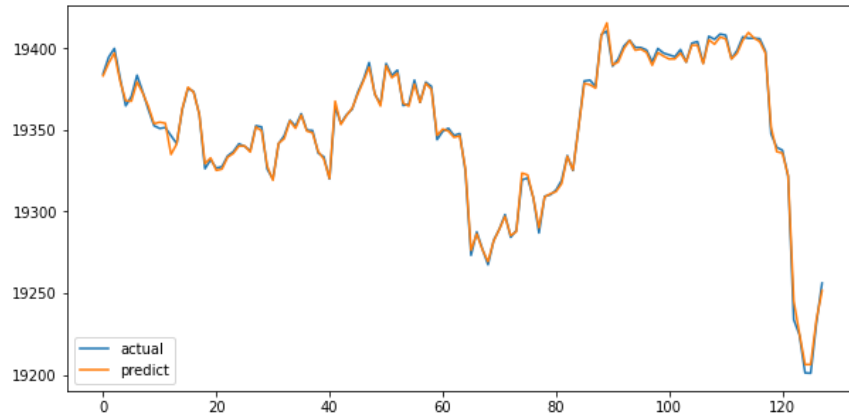
##### 


```

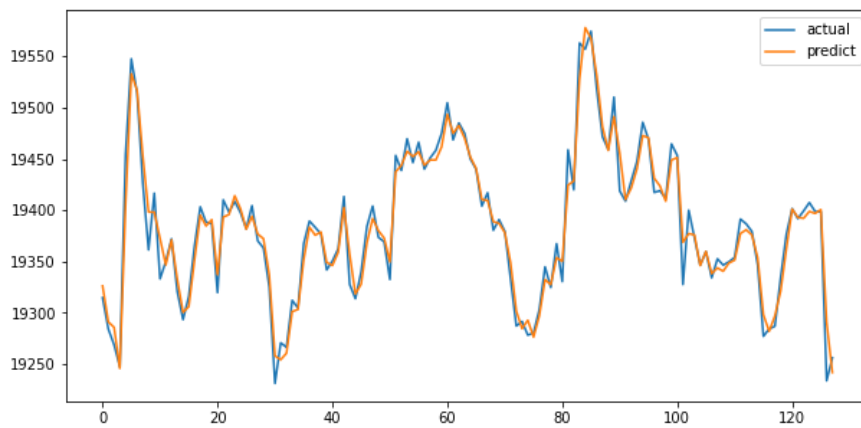
## Lampiran 9 Grafik Pengujian Hasil Pengujian Model

### BTC

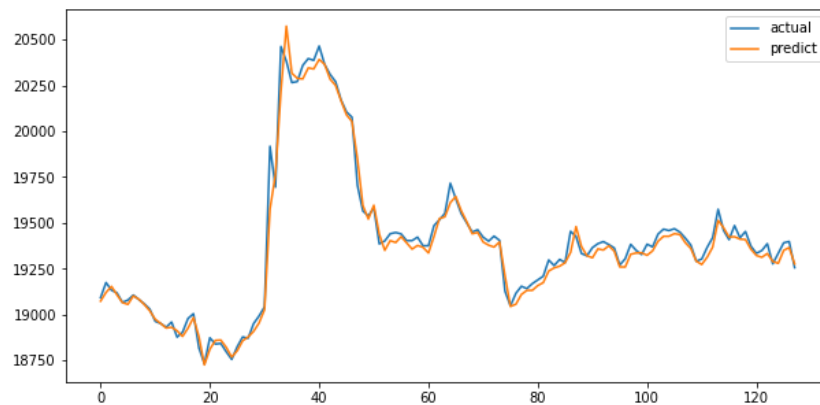
BTC 1m  
Test RMSE: 2.419  
Test Mape: 0.009



BTC 5m  
Test RMSE: 14.733  
Test Mape: 0.055

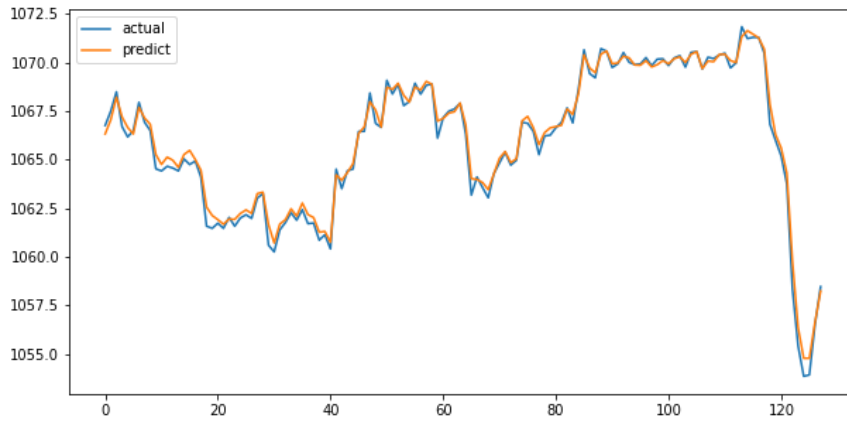


BTC 15m  
Test RMSE: 56.585  
Test Mape: 0.186

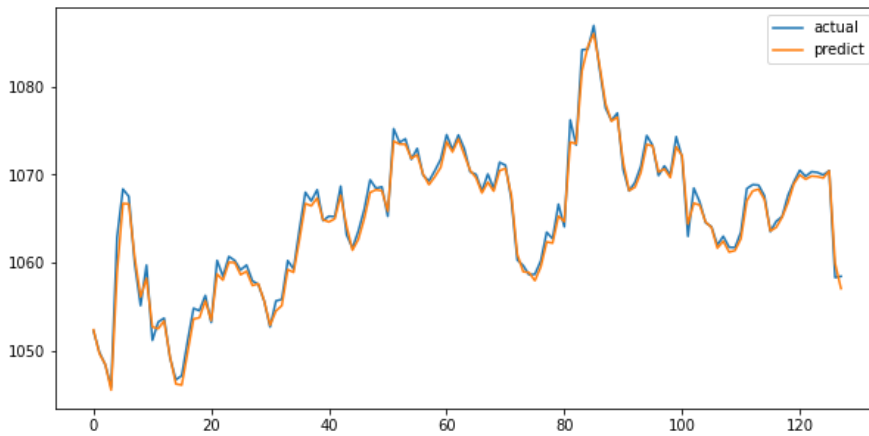


# ETH

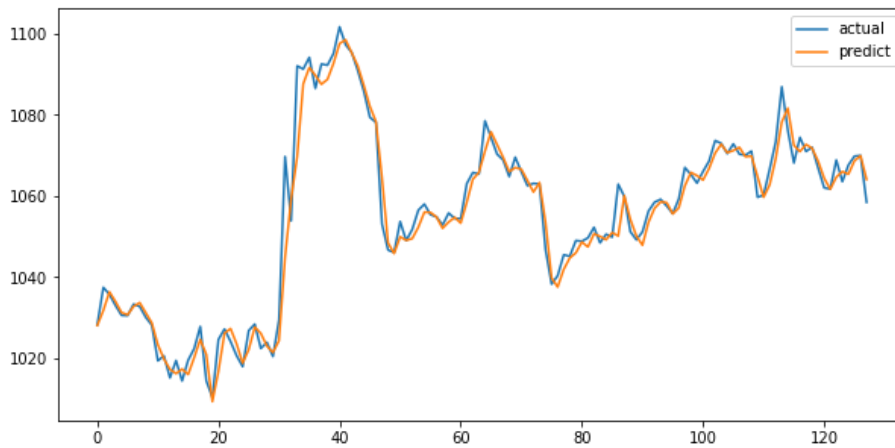
ETH 1m  
Test RMSE: 0.389  
Test Mape: 0.027



ETH 5m  
Test RMSE: 0.876  
Test Mape: 0.063

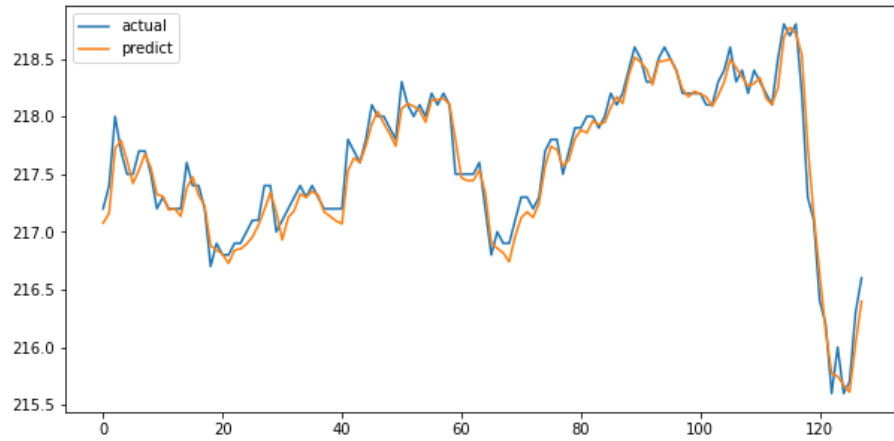


ETH 15m  
Test RMSE: 4.361  
Test Mape: 0.255

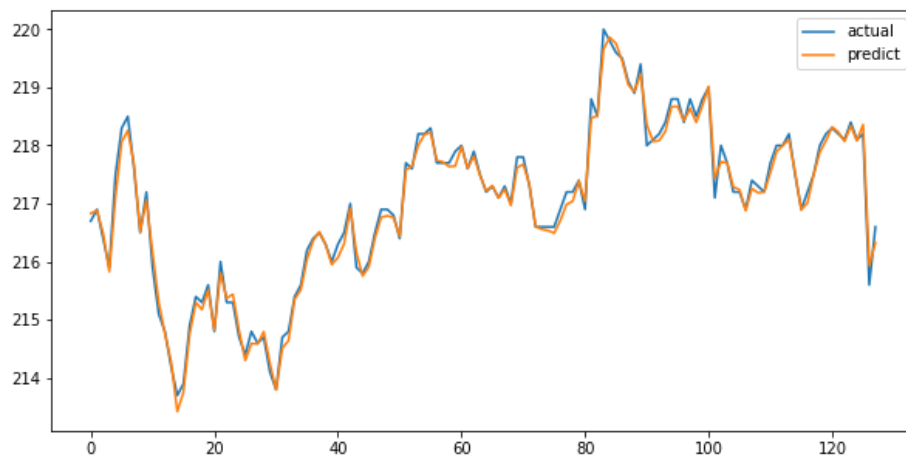


## BNB

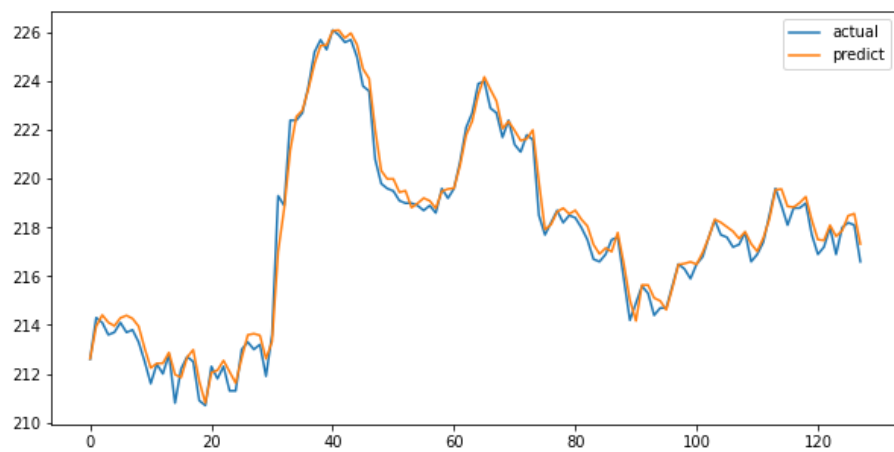
BNB 1m  
Test RMSE: 0.121  
Test Mape: 0.043



BNB 5m  
Test RMSE: 0.140  
Test Mape: 0.049



BNB 15m  
Test RMSE: 0.502  
Test Mape: 0.177

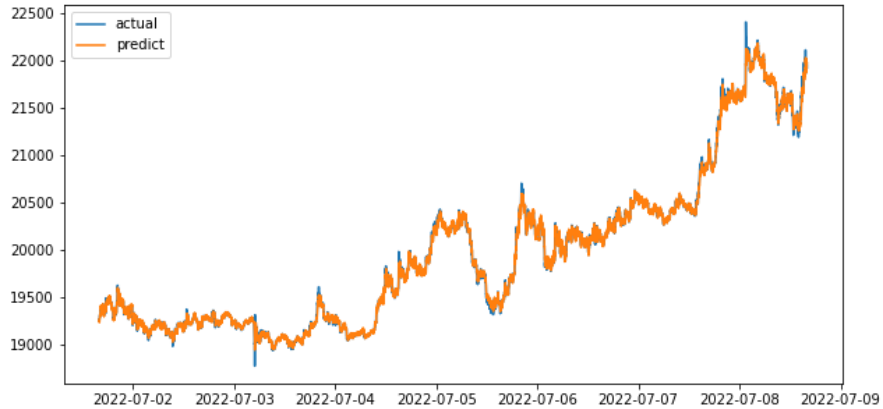




## Lampiran 10 Grafik Pengujian Realtime

### BTC

[LSTM] BTC timeframe 1m 7 Days  
RMSE : 27.548  
MAPE : 0.084



[LSTM] BTC timeframe 5m 7 Days  
RMSE : 40.579  
MAPE : 0.133



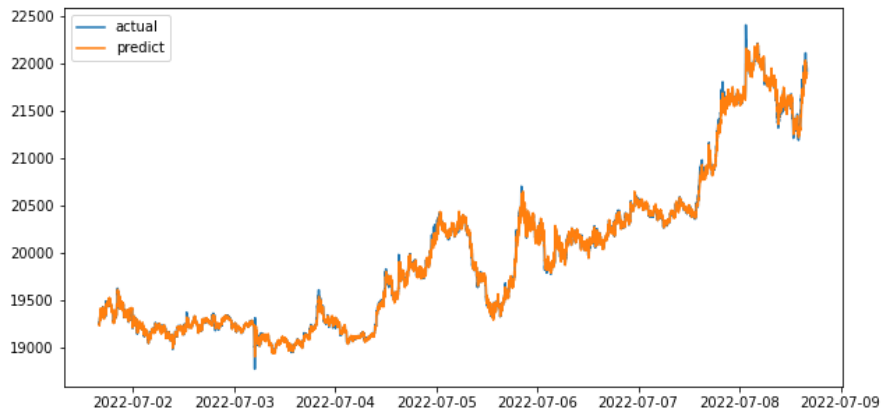
[LSTM] BTC timeframe 15m 7 Days  
RMSE : 61.778  
MAPE : 0.206



[GRU] BTC timeframe 15m 7 Days  
RMSE : 78.772  
MAPE : 0.276



[GRU] BTC timeframe 1m 7 Days  
RMSE : 31.093  
MAPE : 0.096



[GRU] BTC timeframe 5m 7 Days  
RMSE : 53.597  
MAPE : 0.175

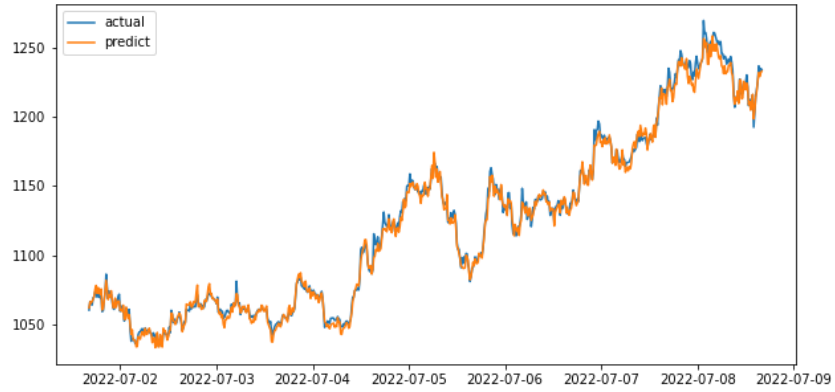


# ETH

[LSTM] ETH timeframe 15m 7 Days

RMSE : 4.106

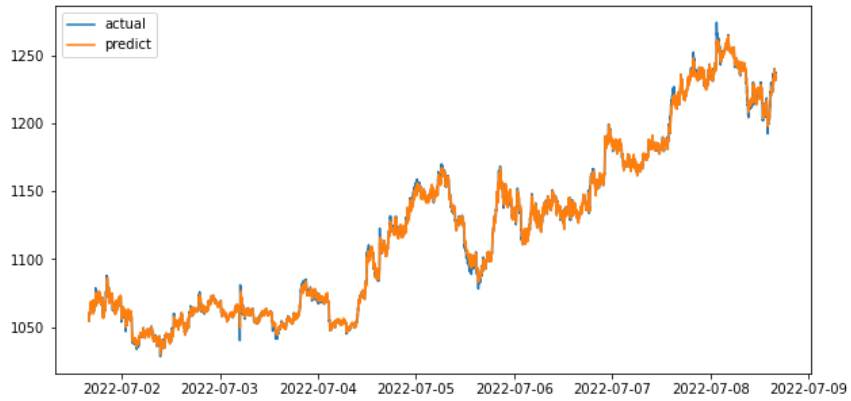
MAPE : 0.274



[LSTM] ETH timeframe 1m 7 Days

RMSE : 1.839

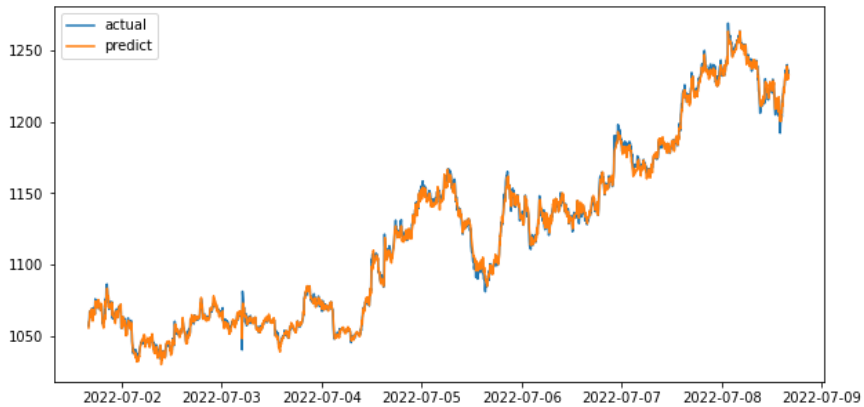
MAPE : 0.112



[LSTM] ETH timeframe 5m 7 Days

RMSE : 2.521

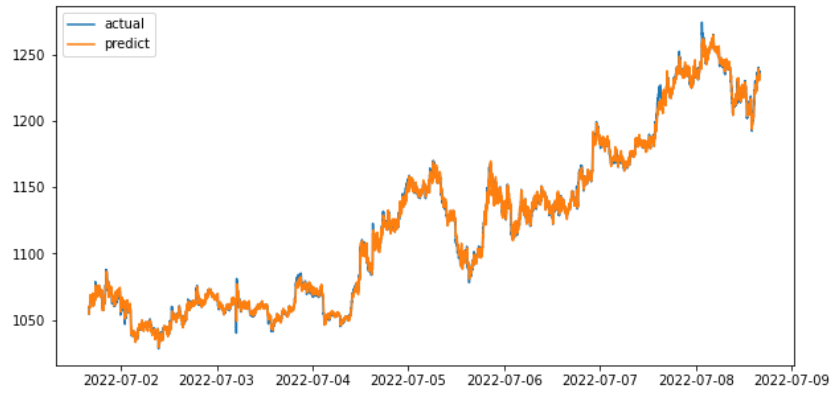
MAPE : 0.166



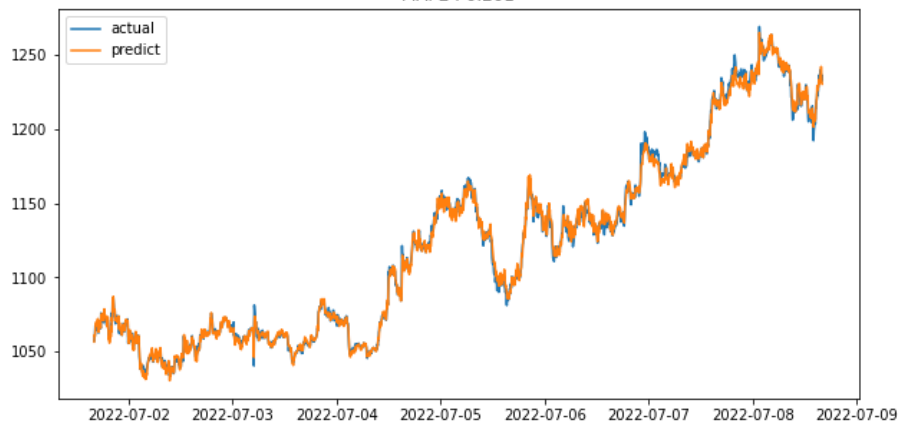
[GRU] ETH timeframe 15m 7 Days  
RMSE : 5.989  
MAPE : 0.392



[GRU] ETH timeframe 1m 7 Days  
RMSE : 2.220  
MAPE : 0.133



[GRU] ETH timeframe 5m 7 Days  
RMSE : 3.080  
MAPE : 0.201



# BNB

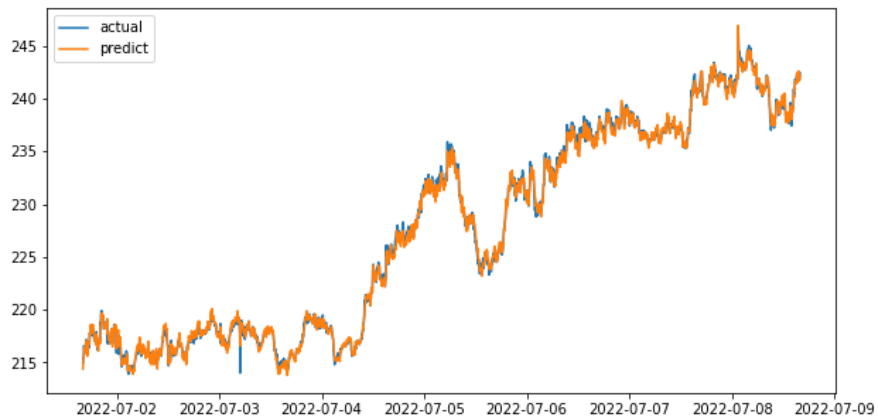
[LSTM] BNB timeframe 15m 7 Days  
RMSE : 0.538  
MAPE : 0.184



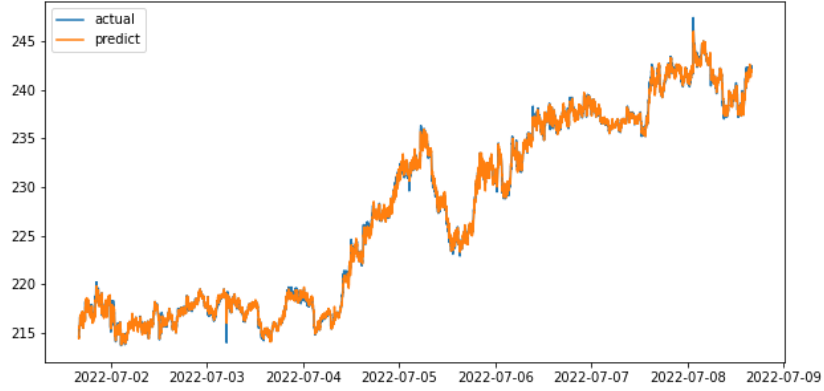
[LSTM] BNB timeframe 1m 7 Days  
RMSE : 0.270  
MAPE : 0.083



[LSTM] BNB timeframe 5m 7 Days  
RMSE : 0.391  
MAPE : 0.129



[GRU] BNB timeframe 1m 7 Days  
RMSE : 0.300  
MAPE : 0.094



[GRU] BNB timeframe 5m 7 Days  
RMSE : 0.486  
MAPE : 0.156



[GRU] BNB timeframe 15m 7 Days  
RMSE : 0.837  
MAPE : 0.287



Lampiran 11 *Riwayat Transaksi*

**BTC**

2022-07-21 18:41:52	BTCBUSD Perpetual	Sell	22,943.3	23.0 BUSD	0.00688299 BUSD	0.02540000 BUSD
2022-07-21 17:24:15	BTCBUSD Perpetual	Buy	22,917.9	23.0 BUSD	0.00687537 BUSD	0.00000000 BUSD
2022-07-21 18:53:25	BTCBUSD Perpetual	Buy	22,963.2	23.0 BUSD	0.00688896 BUSD	-0.02130000 BUSD
2022-07-21 18:42:01	BTCBUSD Perpetual	Sell	22,941.9	23.0 BUSD	0.00688257 BUSD	0.00000000 BUSD
2022-07-21 19:19:58	BTCBUSD Perpetual	Sell	22,726.0	22.8 BUSD	0.00681780 BUSD	-0.13470000 BUSD
2022-07-21 18:59:34	BTCBUSD Perpetual	Buy	22,860.7	22.9 BUSD	0.00685821 BUSD	0.00000000 BUSD
2022-07-21 21:27:40	BTCBUSD Perpetual	Sell	22,695.4	22.7 BUSD	0.00680862 BUSD	-0.06160000 BUSD
2022-07-21 20:23:31	BTCBUSD Perpetual	Buy	22,757.0	22.8 BUSD	0.00273084 BUSD	0.00000000 BUSD
2022-07-22 06:04:32	BTCBUSD Perpetual	Sell	23,145.7	23.2 BUSD	0.00694371 BUSD	0.14570000 BUSD
2022-07-22 00:49:07	BTCBUSD Perpetual	Buy	23,000.0	23.0 BUSD	0.00690000 BUSD	0.00000000 BUSD
2022-07-21 19:30:17	BTCBUSD Perpetual	Buy	22,552.9	22.6 BUSD	0.00676587 BUSD	0.18700000 BUSD
2022-07-21 19:20:23	BTCBUSD Perpetual	Sell	22,739.9	22.8 BUSD	0.00682197 BUSD	0.00000000 BUSD
2022-07-22 07:09:37	BTCBUSD Perpetual	Sell	23,035.5	23.1 BUSD	0.00691065 BUSD	0.03570000 BUSD
2022-07-22 06:54:11	BTCBUSD Perpetual	Buy	22,999.8	23.0 BUSD	0.00689994 BUSD	0.00000000 BUSD
2022-07-22 11:00:22	BTCBUSD Perpetual	Buy	22,994.8	23.0 BUSD	0.00689844 BUSD	0.05790000 BUSD
2022-07-22 10:20:38	BTCBUSD Perpetual	Sell	23,052.7	23.1 BUSD	0.00691581 BUSD	0.00000000 BUSD

2022-07-22 11:28:39	BTCBUSD Perpetual	Buy	22,889.1	22.9 BUSD	0.00686673 BUSD	0.13510000 BUSD
2022-07-22 11:15:01	BTCBUSD Perpetual	Sell	23,024.2	23.1 BUSD	0.00690726 BUSD	0.00000000 BUSD
2022-07-21 20:20:49	BTCBUSD Perpetual	Sell	22,733.5	22.8 BUSD	0.00682005 BUSD	0.12770000 BUSD
2022-07-21 19:30:47	BTCBUSD Perpetual	Buy	22,605.8	22.7 BUSD	0.00678174 BUSD	0.00000000 BUSD

## ETH

2022-07-20 19:54:50	ETHBUSD Perpetual	Buy	1,600.84	140.88 BUSD	0.04226217 BUSD	-0.96800000 BUSD
2022-07-20 19:48:55	ETHBUSD Perpetual	Sell	1,589.84	139.91 BUSD	0.04197177 BUSD	0.00000000 BUSD
2022-07-20 19:41:19	ETHBUSD Perpetual	Sell	1,592.83	138.58 BUSD	0.04157286 BUSD	-0.08874000 BUSD
2022-07-20 19:39:30	ETHBUSD Perpetual	Buy	1,593.85	132.29 BUSD	0.03968686 BUSD	0.00000000 BUSD
2022-07-20 19:29:11	ETHBUSD Perpetual	Buy	1,589.46	127.16 BUSD	0.03814704 BUSD	0.19040000 BUSD
2022-07-20 19:29:11	ETHBUSD Perpetual	Buy	1,589.46	12.72 BUSD	0.00381470 BUSD	0.01904000 BUSD
2022-07-20 19:28:15	ETHBUSD Perpetual	Sell	1,591.84	7.96 BUSD	0.00238776 BUSD	0.00000000 BUSD
2022-07-20 19:28:15	ETHBUSD Perpetual	Sell	1,591.84	6.37 BUSD	0.00191020 BUSD	0.00000000 BUSD
2022-07-20 19:28:15	ETHBUSD Perpetual	Sell	1,591.84	20.70 BUSD	0.00620817 BUSD	0.00000000 BUSD
2022-07-20 19:28:15	ETHBUSD Perpetual	Sell	1,591.84	105.07 BUSD	0.03151843 BUSD	0.00000000 BUSD
2022-07-20 19:18:08	ETHBUSD Perpetual	Buy	1,594.03	39.86 BUSD	0.01195522 BUSD	0.07550000 BUSD
2022-07-20 19:13:42	ETHBUSD Perpetual	Sell	1,597.05	39.93 BUSD	0.01197787 BUSD	0.00000000 BUSD
2022-07-20 19:12:34	ETHBUSD Perpetual	Buy	1,603.41	19.25 BUSD	0.00577227 BUSD	0.02472000 BUSD
2022-07-20 19:11:12	ETHBUSD Perpetual	Sell	1,605.47	19.27 BUSD	0.00577969 BUSD	0.00000000 BUSD



2022-07-20 19:32:52	ETHBUSD Perpetual	Sell	1,592.44	140.14 BUSD	0.04204041 BUSD	0.30536000 BUSD
2022-07-20 19:29:28	ETHBUSD Perpetual	Buy	1,588.97	139.83 BUSD	0.04194880 BUSD	0.00000000 BUSD
2022-07-20 19:32:52	ETHBUSD Perpetual	Sell	1,592.44	140.14 BUSD	0.04204041 BUSD	0.30536000 BUSD
2022-07-20 19:29:28	ETHBUSD Perpetual	Buy	1,588.97	139.83 BUSD	0.04194880 BUSD	0.00000000 BUSD
2022-07-20 19:10:07	ETHBUSD Perpetual	Sell	1,605.15	19.27 BUSD	0.00577854 BUSD	0.13068000 BUSD
2022-07-20 19:01:15	ETHBUSD Perpetual	Buy	1,594.26	6.38 BUSD	0.00191311 BUSD	0.00000000 BUSD
2022-07-21 06:19:47	ETHBUSD Perpetual	Sell	1,543.78	961.78 BUSD	0.28853248 BUSD	0.95319000 BUSD
2022-07-21 06:18:52	ETHBUSD Perpetual	Buy	1,542.25	960.83 BUSD	0.28824652 BUSD	0.00000000 BUSD
2022-07-20 19:32:52	ETHBUSD Perpetual	Sell	1,592.44	140.14 BUSD	0.04204041 BUSD	0.30536000 BUSD
2022-07-20 19:29:28	ETHBUSD Perpetual	Buy	1,588.97	139.83 BUSD	0.04194880 BUSD	0.00000000 BUSD

## BNB

2022-07-22 13:12:46	BNBBUSD Perpetual	Sell	263.660	21.10 BUSD	0.00632784 BUSD	-0.08240000 BUSD
2022-07-22 13:08:12	BNBBUSD Perpetual	Buy	264.690	21.18 BUSD	0.00635256 BUSD	0.00000000 BUSD
2022-07-22 16:43:57	BNBBUSD Perpetual	Buy	267.160	18.71 BUSD	0.00561036 BUSD	0.03500000 BUSD
2022-07-22 16:40:40	BNBBUSD Perpetual	Sell	267.660	18.74 BUSD	0.00562086 BUSD	0.00000000 BUSD
2022-07-22 16:39:35	BNBBUSD Perpetual	Sell	267.730	18.75 BUSD	0.00562233 BUSD	0.02660000 BUSD
2022-07-22 14:10:58	BNBBUSD Perpetual	Buy	267.350	18.72 BUSD	0.00224574 BUSD	0.00000000 BUSD
2022-07-22 14:10:22	BNBBUSD Perpetual	Buy	267.470	18.73 BUSD	0.00561687 BUSD	0.00070000 BUSD
2022-07-22 14:07:08	BNBBUSD Perpetual	Sell	267.480	18.73 BUSD	0.00561708 BUSD	0.00000000 BUSD

2022-07-22 14:05:51	BNBBUSD Perpetual	Sell	267.370	18.72 BUSD	0.00561477 BUSD	0.01750000 BUSD
2022-07-22 14:01:31	BNBBUSD Perpetual	Buy	267.120	18.70 BUSD	0.00560952 BUSD	0.00000000 BUSD
2022-07-22 13:58:07	BNBBUSD Perpetual	Sell	266.440	18.66 BUSD	0.00559524 BUSD	0.02450000 BUSD
2022-07-22 13:55:33	BNBBUSD Perpetual	Buy	266.090	18.63 BUSD	0.00558789 BUSD	0.00000000 BUSD
2022-07-22 13:46:02	BNBBUSD Perpetual	Sell	265.540	21.25 BUSD	0.00637296 BUSD	-0.01600000 BUSD
2022-07-22 13:39:01	BNBBUSD Perpetual	Buy	265.740	21.26 BUSD	0.00637776 BUSD	0.00000000 BUSD
2022-07-22 13:33:44	BNBBUSD Perpetual	Sell	265.550	21.25 BUSD	0.00637320 BUSD	0.04560000 BUSD
2022-07-22 13:26:12	BNBBUSD Perpetual	Buy	264.980	21.20 BUSD	0.00635952 BUSD	0.00000000 BUSD
2022-07-22 13:24:46	BNBBUSD Perpetual	Sell	264.910	21.20 BUSD	0.00635784 BUSD	0.05060000 BUSD
2022-07-22 13:17:54	BNBBUSD Perpetual	Buy	264.280	15.86 BUSD	0.00475703 BUSD	0.00000000 BUSD

# LEMBAR PERBAIKAN SKRIPSI

**“PREDIKSI NILAI TUKAR *CRPTOCURRENCY* JANGKA PENDEK  
DENGAN MENGGUNAKAN *LONG SHORT TERM MEMORY (LSTM)*”**

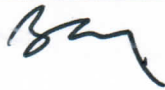



**OLEH:**

**MUHAMMAD FANDLY FADLURACHMAN  
D121181701**


Skripsi ini telah dipertahankan pada Ujian Akhir Sarjana tanggal 26 Oktober 2022.

Telah dilakukan perbaikan penulisan dan isi skripsi berdasarkan usulan dari penguji dan pembimbing skripsi.

Persetujuan perbaikan oleh tim penguji:

	Nama	Tanda Tangan
Ketua	Dr. Ir. Indrabayu, M.T., M.Bus.Sys.,IPM	
Sekretaris	A. Ais Prayogi Alimuddin, S.T., M.Eng.	
Anggota	Hasniaty A, S.T., M.T.,Ph.D.	
	Elly Warni, S.T., M.T.	

Persetujuan Perbaikan oleh pembimbing:

Pembimbing	Nama	Tanda Tangan
I	Dr. Ir.Indrabayu, M.T., M.Bus.Sys.,IPM	
II	A. Ais Prayogi Alimuddin, S.T., M.Eng.	