

## Model Development Phase Template

Date	03 October 2024
Team ID	LTVIP2024TMID24963
Project Title	Time Series Analysis For Bitcoin Price Prediction Using Prophet
Maximum Marks	4 Marks

### Initial Model Training Code, Model Validation and Evaluation Report

The initial model training code will be showcased in the future through a screenshot. The model validation and evaluation report will include classification reports, accuracy, and confusion matrices for multiple models, presented through respective screenshots.

### Initial Model Training Code:

```
from fbprophet.diagnostics import performance_metrics
from fbprophet.plot import plot_cross_validation_metric
from fbprophet.diagnostics import cross_validation

# Assuming 'model' is your fitted Prophet model
df_cv = cross_validation(m, initial='365 days', period='180 days', horizon='365 days')

# Compute performance metrics
df_p = performance_metrics(df_cv)
print(df_p.head())

# Visualize performance metrics
fig = plot_cross_validation_metric(df_cv, metric='mae')

INFO:fbprophet:Making 12 forecasts with cutoffs between 2017-06-08 00:00:00 and 2022-11-09 00:00:00
WARNING:fbprophet:Seasonality has period of 365.25 days which is larger than initial window. Consider increasing initial.

Could not render content for 'application/vnd.jupyter.widget-view+json'
('model_id': 'edc28ab6e3d14773abe5a24d0151dd39', 'version_major': 2, 'version_minor': 0)

horizon    mse    rmse    mae    mape    mdape    coverage
0 37 days $30,460,837.99 $5,519.13 $4,190.97 $0.36 $0.24 $0.15
1 38 days $31,859,113.54 $5,644.39 $4,259.57 $0.37 $0.25 $0.15
2 39 days $33,694,115.73 $5,804.66 $4,340.80 $0.37 $0.25 $0.15
3 40 days $35,482,365.48 $5,956.71 $4,417.51 $0.38 $0.26 $0.15
4 41 days $37,429,333.02 $6,117.95 $4,489.15 $0.38 $0.27 $0.15
```

```
from datetime import datetime
from sklearn.metrics import mean_squared_error, mean_absolute_error, r2_score
import numpy as np

# Assuming 'data_test' is your DataFrame containing true values and 'forecast' is the DataFrame with predicted values
y_true = df1['y']
forecast_before_nov_9 = forecast[forecast['ds'] <= datetime(2023, 11, 9)]

# Extract yhat column from the filtered DataFrame
y_pred = forecast_before_nov_9['yhat']

# Mean Squared Error (MSE)
mse = mean_squared_error(y_true=y_true, y_pred=y_pred)
print(f"Mean Squared Error (MSE): {mse}")

# Mean Absolute Error (MAE)
mae = mean_absolute_error(y_true=y_true, y_pred=y_pred)
print(f"Mean Absolute Error (MAE): {mae}")

# R-squared (R2)
r2 = r2_score(y_true=y_true, y_pred=y_pred)
print(f"R-squared (R2): {r2}")

# Root Mean Squared Error (RMSE)
rmse = np.sqrt(mse)
print(f"Root Mean Squared Error (RMSE): {rmse}")
```

### Model Validation and Evaluation Report:

<b>Model</b>	Mean Squared Error (MSE):	Mean Absolute Error (MAE)	<b>Root Mean Square Error (RMSE)</b>
Prophet	<b>8006516.28151861</b>	<b>1897.6986625308068</b>	<b>2829.578816982946</b>

