Prophet – structural time series model

Prophet = Trend + Seasonality + Holiday + error

$$y(t) = g(t) + s(t) + h(t) + \epsilon$$

- Trend models non periodic changes in the value of the time series.
- Seasonality is the periodic changes like daily, weekly, or yearly seasonality.
- Holiday effects the time series which occur on irregular schedules over a day or a period of days.
- Error terms is what is not explained by the model.

Trend Model

$$g(t) = \frac{C(t)}{1 + \exp(-k(t)(t-m))}$$

- Borrow the concept of population growth model
- C(t) is a time varing carry capacity
- k(t) is a time varing growth rate that has several changing points
- m is the offset parameter

Seasonal Model

$$s(t) = \sum_{n=1}^{N} \left(a_n \cos\left(\frac{2\pi nt}{P}\right) + b_n \sin\left(\frac{2\pi nt}{P}\right) \right)$$

- It is based on a standard Fourior Series.
- The parameter are selected by model performance, such as AIC
- For yearly and weekly seasonality we have found N = 10 and N = 3 respectively to work well for most problems.

Holiday and Event Model

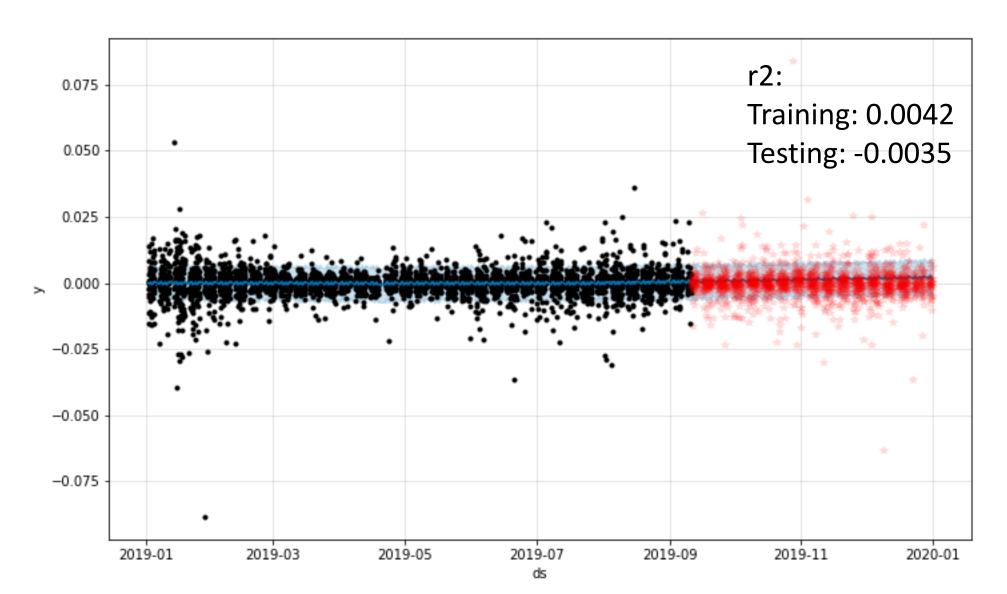
$$Z(t) = [\mathbf{1}(t \in D_1), ..., \mathbf{1}(t \in D_L)]$$
$$h(t) = Z(t)\kappa$$

- by assuming that the effects of holidays are independent, it uses indicator function (whether the date is holiday or not) to model the time series.
- κ is a prior to adjust for seasonality
- For days around the holiday, it labels them as a holiday itself.

Advantages

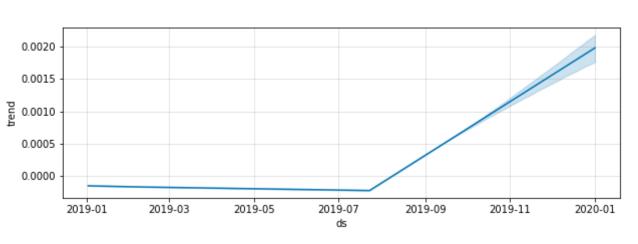
- Decompose easily and Fitting of the model is fast
- Flexibility: Accommodates seasonality with multiple periods
- Intuitive hyper parameters which are easy to tune
- Easy to interpret the model

Result – Final Model

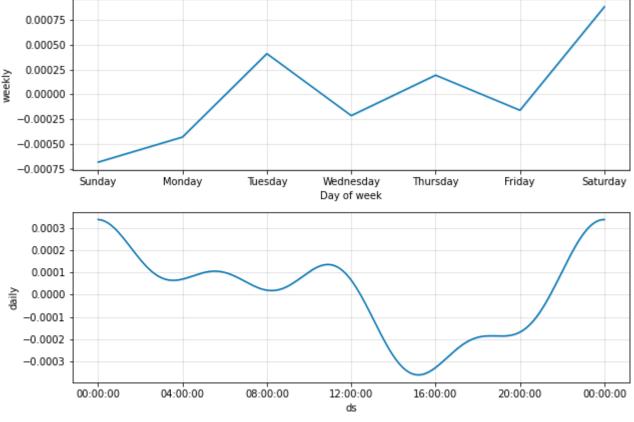


Result – Final Model

Trend



Seasonality (weekly & daily)



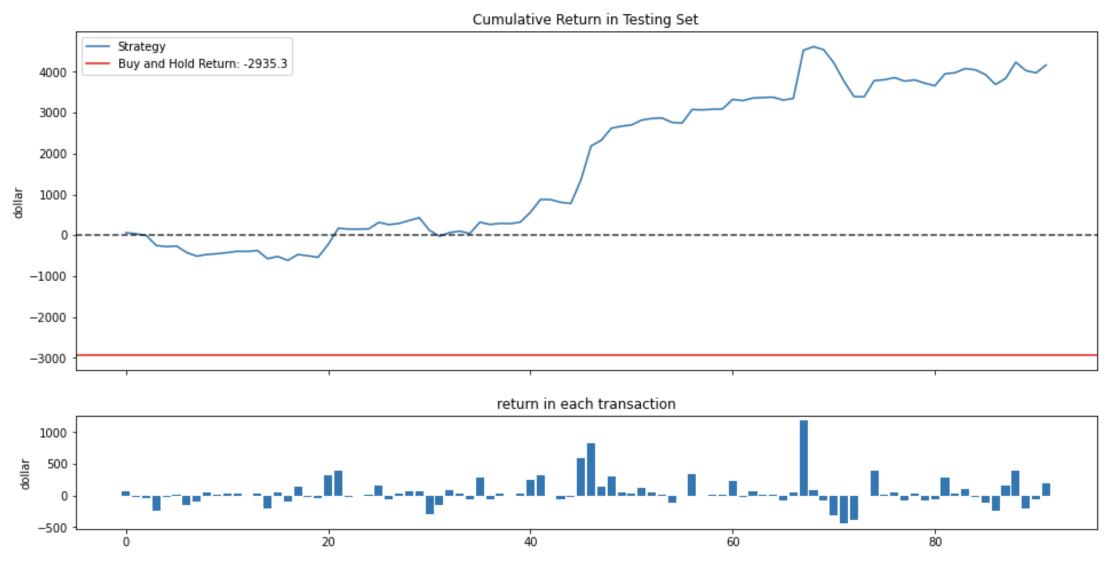
Strategy

- Long and Short strategy:
 - Long: when three consecutive prediction of log return are larger than zero
 - Short: when three consecutive prediction of log return are smaller than zero
 - Hold: otherwise

• If the long signal appears, we buy one unit of BTC. And if the short signal appears, we sell one unit of BTC.

• Our deposit starts with 10,000.

Performance in Testing Set — Rate of Return: 41.7%



Conclusion

• The performance of prophet on log return wasn't ideal, but it can still make profitable strategy after doing some adjustment.

 The result may be different if we take tax and trading cost into account.

• Prophet is good at modelling time series with strong seasonality and with critical changing points. Prophet might perform better on market Index, which has an apparent trend.

Reference

- Paper: https://peerj.com/preprints/3190/
- Package: https://facebook.github.io/prophet/
- Medium blog: https://towardsdatascience.com/time-series-prediction-using-prophet-in-python-35d65f626236