

# Time series analysis using Prophet

Based on the Generalized Additive Models (GAMs)

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

- ▶  $g(t)$ : Trend function (non-periodic changes)
- ▶  $s(t)$ : Seasonality function (periodic changes)
- ▶  $h(t)$ : Holiday effect function
- ▶  $\epsilon_t$ : Error term

# Trend $g(t)$

- ▶ Logistic Trend Model
- ▶ Linear Trend Model

# Logistic Trend Model

Based on the classic logistic growth model

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

- ▶  $C$ : Carrying capacity
- ▶  $k$ : Growth rate
- ▶  $m$ : Offset parameter

# Logistic Trend Model

Introduction of changepoints  $s_j$ , where  $j \in S$ : timestamps at which growth rate can change

- ▶  $k + \sum_{j|t \geq s_j} \delta_j$ : growth rate at time  $t$

Prophet utilizes several vectorial representations

- ▶  $\mathbf{a}(t) \in \{0, 1\}^S$
- ▶  $a_j(t) = \begin{cases} 1 & \text{if } t \geq s_j \\ 0 & \text{otherwise} \end{cases}$
- ▶  $k + \mathbf{a}(t)^T \boldsymbol{\delta}$ : growth rate at time  $t$

# Logistic Trend Model

Phophet changes to the classical equation

$$g(t) = \frac{C(t)}{1 + \exp(-(k + \mathbf{a}(t)^T \boldsymbol{\delta})(t - (m + \mathbf{a}(t)^T \boldsymbol{\gamma})))}$$

- ▶  $C \rightarrow C(t)$ : Carrying capacity changes over time
- ▶  $k \rightarrow k + \mathbf{a}(t)^T \boldsymbol{\delta}$ : Growth rate can change over time at specific changepoints  $s_j$
- ▶  $m \rightarrow m + \mathbf{a}(t)^T \boldsymbol{\gamma}$ : Adjustment to the offset parameter to account for changepoints

# Linear Trend Model

$$g(t) = \underbrace{(k + \mathbf{a}(t)^T \boldsymbol{\delta})}_{\text{slope}} t + \underbrace{(m + \mathbf{a}(t)^T \boldsymbol{\gamma})}_{\text{intercept}}$$

- ▶  $k + \mathbf{a}(t)^T \boldsymbol{\delta}$ : Growth rate
- ▶  $m + \mathbf{a}(t)^T \boldsymbol{\gamma}$ : Offset parameter

# Seasonality $s(t)$

- ▶ Additive seasonality:  $g(t) + s(t)$
- ▶ Multiplicative seasonality:  $g(t)s(t)$

# Seasonality $s(t)$

Based on Fourier series cosine-sine form

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

- ▶  $\mathbf{X}(t) = [\cos(\frac{2\pi t}{P}), \sin(\frac{2\pi t}{P}), \dots, \cos(\frac{2\pi Nt}{P}), \sin(\frac{2\pi Nt}{P})]$
- ▶  $\boldsymbol{\beta} = [a_1, b_1, \dots, a_N, b_N]^T$
- ▶  $N$ : Order of Fourier series
- ▶  $P$ : Period



# Holiday $h(t)$

Introducing holidays  $D_i$ , where  $i \in L$ : days at which an holiday occurs

$$h(t) = \mathbf{Z}(t)\kappa$$

- ▶  $\mathbf{Z}(t) = \begin{cases} z_i = 1 & \text{if } i \in L \\ z_i = 0 & \text{otherwise} \end{cases}$
- ▶  $\mathbf{Z}(t)$ : vector of holiday dummies
- ▶  $\kappa$ : Effect of the holiday

# Final Model

$$y|m, \delta, \beta, \kappa, \epsilon \sim \mathcal{N}(g(t) + s(t) + h(t), \epsilon)$$

- ▶ The parameters are obtained as Maximum A Posterior (MAP) estimates

# Additional Features

- ▶ Seasonalities: Custom seasonalities can be added with `add_seasonality()`
- ▶ Regressors: Additional regressors can be added using `add_regressor()`, although their future values have to be known
- ▶ Uncertainty: Confidence intervals are given by Markov chain Monte Carlo sampling

# Practical Implementation

1. Import Prophet
2. Fit the model
3. Tune the parameters
4. Evaluate the results
5. Repeat step 3. and 4. as needed