

Probabilistic Boosting Tree

|  |                    |     |                             |   |          |
|--|--------------------|-----|-----------------------------|---|----------|
| LSS  | GamLSS/GamboostLSS | [1] | LSS location/shape/skewness | R | gamlss   |
| Gaussian Process Regression[2] / Distributional learning [3] |                    |     |                             |   |          |
| Quantile Regression  |                    |     |                             |   | quantile |

Probabilistic Boosting Tree

$y \quad x \quad y \quad p(y|x) \quad \text{boosting tree} \quad T_1 \quad l_i \quad p_{l_i}(y)$

$p_{T_1}(y|x)$

d Boosting

$q(y|x) = \frac{\prod_{i=0}^{d-1} p_{T_i}(y|x)}{C}$

C

$q_{T_m}(y|x) = \frac{p_{T_m}(y|x) \prod_{i=0}^{m-1} p_{T_i}(y|x)}{C}$

$q_{T_{m-1}}(y|x) = \prod_{i=0}^{m-1} p_{T_i}(y|x) \quad p_{T_m}(y|x) \quad q_{T_m}$   
Gamma Gamma 0

$q_{T_m}(y|x) = \frac{1}{\Gamma(k_m)\theta_m^{k_m}} y^{k_m-1} e^{-\frac{y}{\theta_m}}$

$k_{m-1} \rightarrow k_m \quad \theta_{m-1} \rightarrow \theta_m$

$k_m = k_{m-1} + \Delta k \quad \theta_m = \theta_{m-1} + \Delta \theta \quad \Delta k \quad \Delta \theta$

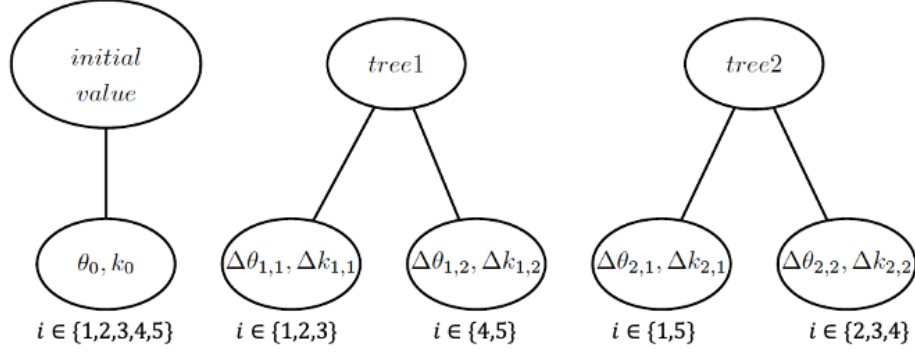


Figure 1: PBTree

- $\theta_{2,1} = \theta_0 + \Delta\theta_{1,1} + \Delta\theta_{2,1}$
- $\theta_{2,2} = \theta_0 + \Delta\theta_{1,1} + \Delta\theta_{2,2}$
- $\theta_{2,5} = \theta_0 + \Delta\theta_{1,2} + \Delta\theta_{2,1}$

n      m

$$L_m = -\log\left(\prod_{i=1}^n p(y_i | \theta_{m-1,i} + \Delta\theta_l, k_{m-1,i} + \Delta k_l)\right), \quad i \in S_{m,l}$$

$$\theta_{m-1,i} = \theta_0 + \sum_{j=0}^{m-1} \theta_l$$

$$k_{m-1,i} = k_0 + \sum_{j=0}^{m-1} k_l$$

$$i \in S_{m,l} \quad i \in m \quad l$$

$$\Delta\theta \quad \Delta k$$

$$\arg \min_{S, \Theta, K} L_m$$

$$L_m = -\log\left(\prod_{i=0}^n p(y_i | \theta_{m-1,i} + \Delta\theta_{l, \{i \in S_{m,l}\}}, k_{m-1,i} + \Delta k_{l, \{i \in S_{m,l}\}})\right) = -\sum_{i=0}^n \log(p(y_i | \theta_{m-1,i} + \Delta\theta_{l, \{i \in S_{m,l}\}}, k_{m-1,i} + \Delta k_{l, \{i \in S_{m,l}\}}))$$

$$m-l \quad (x_i, y_i), i = \{1 \dots s\}$$

$$\frac{\partial L_m}{\partial k_{m,l}} = - \sum_{i=1}^s (\log(\frac{y_i}{\theta_{m-1,l}}) + F(k_{m-1,l}))$$

$$\frac{\partial L_m}{\partial \theta_{m,l}} = - \sum_{i=1}^s (\frac{y_i}{\theta_{m-1,l}^2} - \frac{k_{m-1,l}}{\theta_{m-1,l}})$$

$$\eta_1, \eta_2$$

$$\Delta k_{m,l} = \eta_1 \frac{\partial L_m}{\partial k_{m,l}}$$

$$\Delta \theta_{m,l} = \eta_2 \frac{\partial L_m}{\partial \theta_{m,l}}$$