

Conditional Quantile Regression Article Proposal

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Overview

Let a time series be given by

$$y_t = A(L)y_t + \beta x_t + \varepsilon_t,$$

where the distribution of ε_t is unknown. (...)

Nonparametric model

Nonparametric model - Formulation

$$\min_{q_{\alpha t}, \delta_t^+, \delta_t^-, \xi_t}$$

s.t.

$$\sum_{\alpha \in A} \sum_{t \in T'} \left(\alpha \delta_{t\alpha}^+ + (1 - \alpha) \delta_{t\alpha}^- \right)$$

$$+ \lambda_1 \sum_{t \in T'} \gamma_{t\alpha} + \lambda_2 \sum_{t \in T'} \xi_{t\alpha}$$

$$\delta_t^+ - \delta_{t\alpha}^- = y_t - q_{t\alpha},$$

$$D_{t\alpha}^1 = \frac{q_{\alpha t+1} - q_{\alpha t}}{x_{t+1} - x_t},$$

$$D_{t\alpha}^2 = \frac{\left(\frac{q_{\alpha t+1} - q_{\alpha t}}{x_{t+1} - x_t} \right) - \left(\frac{q_{\alpha t} - q_{\alpha t-1}}{x_t - x_{t-1}} \right)}{x_{t+1} - 2x_t + x_{t-1}}.$$

$$\gamma_{t\alpha} \geq D_{t\alpha}^1,$$

$$\gamma_{t\alpha} \geq -D_{t\alpha}^1,$$

$$\xi_{t\alpha} \geq D_{t\alpha}^2,$$

$$\xi_{t\alpha} \geq -D_{t\alpha}^2,$$

$$\delta_{t\alpha}^+, \delta_{t\alpha}^-, \gamma_{t\alpha}, \xi_{t\alpha} \geq 0,$$

$$q_{t\alpha} \leq q_{t\alpha'},$$

$$\forall t \in T', \forall \alpha \in A,$$

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$$\forall t \in T', \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha' \Rightarrow q_{t\alpha} \leq q_{t\alpha'}$$

Nonparametric vs. Linear Model

- ▶ The nonparametric approach is more flexible to capture heteroscedasticity.

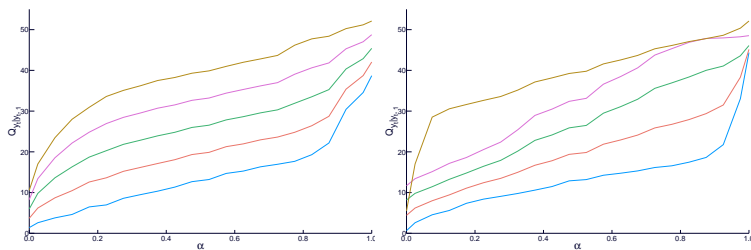


Figure 1: Estimated quantile functions, for different values of y_{t-1} . On the left using a linear model and using a nonparametric approach on the right.

Nonparametric vs. Linear Model

- This flexibility might lead to overfitting, if we don't select a proper penalty, as shown below:

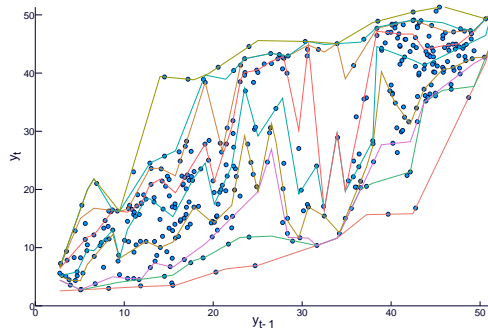


Figure 2: Example of a overfitted quantile function

Linear Models

Formulation

$$\min_{\beta_{0\alpha}, \beta_{\alpha}, \varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^-} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^+ + (1 - \alpha) \varepsilon_{t\alpha}^- \right) \quad (11)$$

$$\text{s.t.} \quad \varepsilon_{t\alpha}^+ - \varepsilon_{t\alpha}^- = y_t - \beta_{0\alpha} - \beta_{\alpha}^T x_t, \quad \forall t \in T, \forall \alpha \in A, \quad (12)$$

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \geq 0, \quad \forall t \in T, \forall \alpha \in A, \quad (13)$$

$$\beta_{0\alpha} + \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t, \quad \forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha', \quad (14)$$

One model for each α -quantile - Resume

In this part, we investigate the usage of MILP to select which variables are included in the model, by using a constraint which limits them to a number of K . This means that only K coefficients $\beta_{p\alpha}$ may have nonzero values, for each α -quantile. This assumption is modeled with binary variables $z_{p\alpha}$, which indicates whether $\beta_{p\alpha}$ is included or not.

One model for each α -quantile - Formulation

$$\min_{\beta_{0\alpha}, \beta_{\alpha}, z_{p\alpha}, \varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^-} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^+ + (1 - \alpha) \varepsilon_{t\alpha}^- \right) \quad (15)$$

$$\text{s.t.} \quad \varepsilon_{t\alpha}^+ - \varepsilon_{t\alpha}^- = y_t - \beta_{0\alpha} - \sum_{p=1}^P \beta_{p\alpha} x_{t,p}, \quad \forall t \in T, \forall \alpha \in A, \quad (16)$$

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \geq 0, \quad \forall t \in T, \forall \alpha \in A, \quad (17)$$

$$-Mz_{p\alpha} \leq \beta_{p\alpha} \leq Mz_{p\alpha}, \quad \forall \alpha \in A, \forall p \in P, \quad (18)$$

$$\sum_{p=1}^P z_{p\alpha} \leq K, \quad \forall \alpha \in A, \quad (19)$$

$$z_{p\alpha} \in \{0, 1\}, \quad \forall \alpha \in A, \forall p \in P, \quad (20)$$

$$\beta_{0\alpha} + \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t, \quad \forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha', \quad (21)$$

Defining groups for α -quantiles - Resume

- ▶ Now, adding groups of quantiles.
- ▶ Each probability α belongs to a group g .
- ▶ The total number of groups is limited to G .
- ▶ Incorporation of new integer variables.
- ▶ Total number of valid solution falls.

Defining groups for α -quantiles - Formulation

$$\min_{\beta_{0\alpha}, \beta_{\alpha}, z_{p\alpha}} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^+ + (1 - \alpha) \varepsilon_{t\alpha}^- \right) \quad (22)$$

$$\text{s.t.} \quad \varepsilon_{t\alpha}^+ - \varepsilon_{t\alpha}^- = y_t - \beta_{0\alpha} - \sum_{p=1}^P \beta_{p\alpha} x_{t,p}, \quad \forall t \in T, \forall \alpha \in A, \quad (23)$$

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \geq 0, \quad \forall t \in T, \forall \alpha \in A, \quad (24)$$

$$-Mz_{p\alpha g} \leq \beta_{p\alpha} \leq Mz_{p\alpha g}, \quad \forall \alpha \in A, \forall p \in P, \forall g \in G \quad (25)$$

$$z_{p\alpha g} := 2 - (1 - z_{pg}) - I_{g\alpha} \quad (26)$$

$$\sum_{p=1}^P z_{pg} \leq K, \quad \forall g \in G, \quad (27)$$

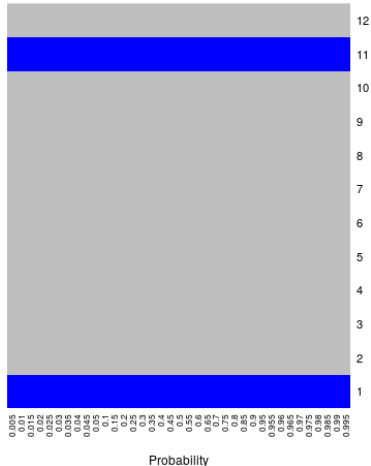
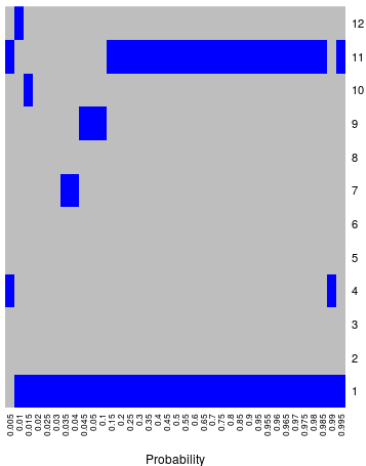
$$\beta_{0\alpha} + \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t, \quad \forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha', \quad (28)$$

$$\sum_{g \in G} I_{g\alpha} = 1, \quad \forall \alpha \in A, \quad (29)$$

$$I_{g\alpha}, z_{pg} \in \{0, 1\}, \quad \forall p \in P, \quad \forall g \in G, \quad (30)$$

Results

- We start to notice, from $K = 2$, that by letting



Defining groups by the introduction of switching variable - Formulation

$$\min_{\beta_{0\alpha}, \beta_{\alpha}, z_{p\alpha}, \varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^-} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^+ + (1 - \alpha) \varepsilon_{t\alpha}^- \right) \quad (31)$$

$$\text{s.t.} \quad \varepsilon_{t\alpha}^+ - \varepsilon_{t\alpha}^- = y_t - \beta_{0\alpha} - \sum_{p=1}^P \beta_{p\alpha} x_{t,p}, \quad \forall t \in T, \forall \alpha \in A, \quad (32)$$

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \geq 0, \quad \forall t \in T, \forall \alpha \in A, \quad (33)$$

$$-Mz_{p\alpha} \leq \beta_{p\alpha} \leq Mz_{p\alpha}, \quad \forall \alpha \in A, \forall p \in \{1, \dots, P\}, \quad (34)$$

$$\sum_{p=1}^P z_{p\alpha} \leq K, \quad \forall \alpha \in A, \quad (35)$$

$$z_{p\alpha} \in \{0, 1\}, \quad \forall \alpha \in A, \forall p \in \{1, \dots, P\}, \quad (36)$$

$$\beta_{0\alpha} + \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t, \quad \forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha', \quad (37)$$

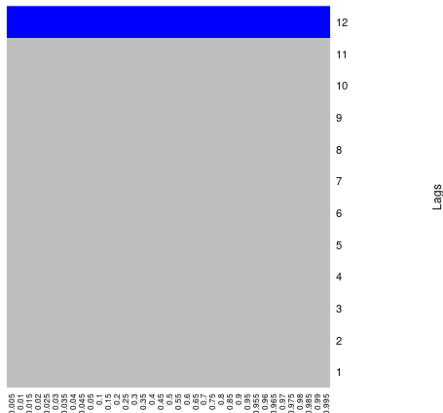
$$z_{p\alpha} - z_{p\alpha+1} \leq m_{p\alpha}, \quad \forall \alpha \in A', \quad \forall p \in P \quad (38)$$

$$\sum_{\alpha \in A'} r_{\alpha} \leq |G| - 1 \quad (39)$$

$$(40)$$

where $A' = A \setminus \{|A|\}$

- ▶ As there are much less possibilities when $K = 1$, every method gets pretty fast to the optimum result. Selecting the 12TH lag was the best choice.



Results

Results