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. In this case, the usage of a parametric model is hardly attrative.

- ▶ Quantile regression is one technique available to model this time series dynamics, by estimating a phin grid of α -quantiles at once and forming a data-driven conditional distribution.
- ▶ We explore different strategies of estimating the Conditional Quantile Regression focused on approaching the conditional distribution or $y_{t+h|t}$, for a given horizon h.

Linear Models

Linear Models

Linear Models - Formulation

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

subject to

$$\varepsilon_{t\alpha}^{+} - \varepsilon_{t\alpha}^{-} = y_t - \beta_{0\alpha} - \beta_{\alpha}^{T} x_t, \forall t \in T, \quad \forall \alpha \in A,$$
 (2)

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \ge 0, \qquad \forall t \in \mathcal{T}, \forall \alpha \in \mathcal{A},$$
 (3)

$$\beta_{0\alpha} + \beta_{\alpha}^{T} x_{t} \leq \beta_{0\alpha'} + \beta_{\alpha'}^{T} x_{t},$$

$$\forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha',$$
(4)

Linear Models - Resume

- As there are many explanatory variables for y_t , it is interesting to do a regularization process in order to select only a subset to compose the model.
- ► The next slides are going to cover a few different ways to achieve this using Mixed Integer Linear Programming.

Linear Models

LM

Regularization by MILP - Resume

- ▶ MILP models allow only K variables to be used for each α -quantile. This means that only K coefficients $\beta_{p\alpha}$ may have nonzero values, for each α -quantile. It must be guaranteed by constraints on the optimization problem.
- We present three forms of grouping probabilities while selecting variables

One model for each α -quantile - Formulation

$$\min_{\beta_{0\alpha},\beta_{\alpha},z_{p\alpha},\varepsilon_{t\alpha}^{+},\varepsilon_{t\alpha}^{-}} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^{+} + (1-\alpha)\varepsilon_{t\alpha}^{-} \right)$$
 (5)

subject to

$$\varepsilon_{t\alpha}^{+} - \varepsilon_{t\alpha}^{-} = y_{t} - \beta_{0\alpha} - \sum_{p=1}^{P} \beta_{p\alpha} x_{t,p},$$

$$\forall t \in T, \forall \alpha \in A,$$
 (6)

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- \ge 0, \qquad \forall t \in T, \forall \alpha \in A,$$
 (7)

$$-Mz_{p\alpha} \le \beta_{p\alpha} \le Mz_{p\alpha}, \qquad \forall \alpha \in A, \forall p \in P,$$
(8)

$$\sum_{n=1}^{\infty} z_{p\alpha} \le K, \qquad \forall \alpha \in A, \tag{9}$$

$$z_{p\alpha} \in \{0,1\},$$
 $\forall \alpha \in A, \forall p \in P,$ (10)

$$\beta_{0\alpha} + \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t,$$

$$\forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha',$$
 (11)

Defining groups for α -quantiles - Formulation

$$\min_{\beta_{0\alpha}, \beta_{\alpha}, z_{p\alpha} \varepsilon_{t\alpha}^{+}, \varepsilon_{t\alpha}^{-}} \sum_{\alpha \in A} \sum_{t \in T} \left(\alpha \varepsilon_{t\alpha}^{+} + (1 - \alpha) \varepsilon_{t\alpha}^{-} \right) \tag{12}$$

s.t
$$\varepsilon_{t\alpha}^+ - \varepsilon_{t\alpha}^- = y_t - \beta_{0\alpha} - \sum_{p=1}^P \beta_{p\alpha} x_{t,p}, \quad \forall t \in T, \forall \alpha \in A,$$
 (13)

$$\varepsilon_{t\alpha}^+, \varepsilon_{t\alpha}^- > 0, \qquad \forall t \in T, \forall \alpha \in A,$$
 (14)

$$-Mz_{p\alpha g} \le \beta_{p\alpha} \le Mz_{p\alpha g}, \qquad \forall \alpha \in A, \forall p \in P, \forall g \in G$$
 (15)

$$z_{p\alpha g} := 2 - (1 - z_{pg}) - I_{g\alpha}$$
 (16)

$$-2 - (1 - 2pg) - ig\alpha \tag{10}$$

$$\sum_{p=1}^{P} z_{pg} \le K, \qquad \forall g \in G, \tag{17}$$

$$\beta_{0\alpha} + \beta_{\alpha}^{\mathsf{T}} x_{t} \leq \beta_{0\alpha'} + \beta_{\alpha'}^{\mathsf{T}} x_{t}, \qquad \forall t \in \mathsf{T}, \forall (\alpha, \alpha') \in \mathsf{A} \times \mathsf{A}, \alpha < \alpha',$$

(18)

$$\sum_{\sigma \in \mathcal{L}} I_{g\alpha} = 1, \qquad \forall \alpha \in A, \tag{19}$$

$$I_{g\alpha}, z_{pg} \in \{0, 1\}, \qquad \forall p \in P, \forall g \in G,$$
 (20)

Defining groups by the introduction of switching variable - Formulation

$$\min_{\beta_{0\alpha},\beta_{\alpha},z_{p\alpha}\varepsilon_{t\alpha}^{+},\varepsilon_{t\alpha}^{-}} \sum_{\alpha\in A} \sum_{t\in T} \left(\alpha\varepsilon_{t\alpha}^{+} + (1-\alpha)\varepsilon_{t\alpha}^{-}\right) \tag{21}$$
s.t
$$\varepsilon_{t\alpha}^{+} - \varepsilon_{t\alpha}^{-} = y_{t} - \beta_{0\alpha} - \sum_{p=1}^{P} \beta_{p\alpha}x_{t,p}, \qquad \forall t\in T, \forall \alpha\in A, \tag{22}$$

$$\varepsilon_{t\alpha}^{+}, \varepsilon_{t\alpha}^{-} \geq 0, \qquad \forall t\in T, \forall \alpha\in A, \tag{23}$$

$$-Mz_{p\alpha} \leq \beta_{p\alpha} \leq Mz_{p\alpha}, \qquad \forall \alpha\in A, \forall p\in \{1,\ldots,P\}, \tag{24}$$

$$\sum_{p=1}^{P} z_{p\alpha} \leq K, \qquad \forall \alpha\in A, \tag{25}$$

$$\begin{aligned} z_{p\alpha} &\in \{0, 1\}, \\ \beta_{0\alpha} &+ \beta_{\alpha}^T x_t \leq \beta_{0\alpha'} + \beta_{\alpha'}^T x_t, \end{aligned}$$

 $z_{p\alpha} - z_{p\alpha+1} \leq m_{p\alpha}$

 $\sum_{\alpha \in A'} r_{\alpha} \leq |G| - 1$

 $\forall t \in T, \forall (\alpha, \alpha') \in A \times A, \alpha < \alpha',$

 $\forall \alpha \in A, \forall p \in \{1, \dots, P\},\$

$$\forall \alpha \in A', \quad \forall p \in P \tag{28}$$

(30)

(26)

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

Nonparametric model

Nonparametric model

Nonparametric model - Formulation

$$\begin{aligned} & \underset{q_{\alpha t}, \delta_{t}^{+}, \delta_{t}^{-}, \xi_{t}}{\min} & \sum_{\alpha \in A} \sum_{t \in T'} \left(\alpha \delta_{t \alpha}^{+} + (1 - \alpha) \delta_{t \alpha}^{-}\right) \\ & + \lambda_{1} \sum_{t \in T'} \gamma_{t \alpha} + \lambda_{2} \sum_{t \in T'} \xi_{t \alpha} \\ & s.t. & \delta_{t}^{+} - \delta_{t \alpha}^{-} = y_{t} - q_{t \alpha}, & \forall t \in T', \forall \alpha \in A, \\ & D_{t \alpha}^{1} = \frac{q_{\alpha t + 1} - q_{\alpha t}}{x_{t + 1} - x_{t}}, & \forall t \in T', \forall \alpha \in A, \\ & D_{t \alpha}^{2} = \frac{\left(\frac{q_{\alpha t + 1} - q_{\alpha t}}{x_{t + 1} - x_{t}}\right) - \left(\frac{q_{\alpha t} - q_{\alpha t - 1}}{x_{t} - x_{t - 1}}\right)}{x_{t + 1} - 2x_{t} + x_{t - 1}}. & \forall t \in T', \forall \alpha \in A, \\ & \gamma_{t \alpha} \geq D_{t \alpha}^{1}, & \forall t \in T', \forall \alpha \in A, \\ & \gamma_{t \alpha} \geq D_{t \alpha}^{1}, & \forall t \in T', \forall \alpha \in A, \\ & \xi_{t \alpha} \geq D_{t \alpha}^{2}, & \forall t \in T', \forall \alpha \in A, \\ & \xi_{t \alpha} \geq D_{t \alpha}^{2}, & \forall t \in T', \forall \alpha \in A, \\ & \xi_{t \alpha} \geq D_{t \alpha}^{2}, & \forall t \in T', \forall \alpha \in A, \\ & \delta_{t \alpha}^{+}, \delta_{t \alpha}^{-}, \gamma_{t \alpha}, \xi_{t \alpha} \geq 0, & \forall t \in T', \forall \alpha \in A, \\ & q_{t \alpha} \leq q_{t \alpha'}, & \forall t \in T', \forall \alpha \in A, \end{aligned}$$

Nonparametric vs. Linear Model

► The nonparametric approach is more flexible to capture heteroscedasticity.

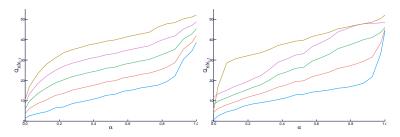


Figure 1: Estimated quantile functions, for different values of y_{t-1} . On the left using a linear model and using a nonparametric approach on the right.

Nonparametric vs. Linear Model

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

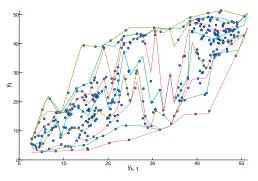


Figure 2: Example of a overfitted quantile function

Goals

Goals

Goals of this work are:

- Using multiple quantiles to estimate the empirical conditional distribution of variables in a time series
- Producing a model identification methodology
- Natural Ressources and Financial applications