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**Optimal Forecast Reconciliation for Hierarchical and Grouped  
Time Series Through Trace Minimization  
Wickramasuriya, Athanasopoulos & Hyndman, (2019)**

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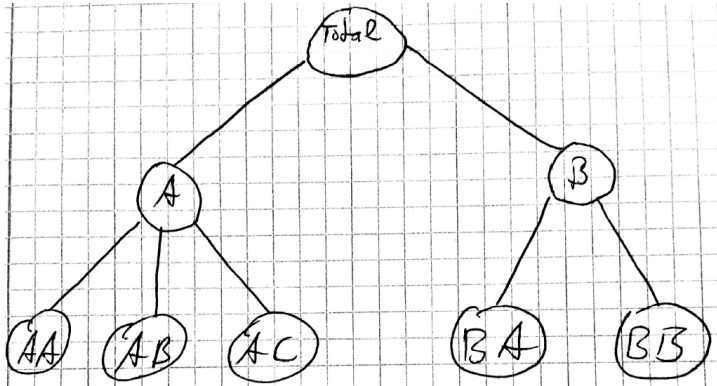
## Contents

**Problem Statement**

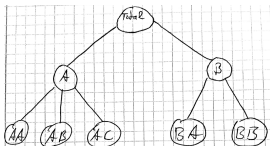
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**Figure:** An easy example of a hierarchy. (Recreated from [Wickramasuriya et al., 2017](#)). AA, AB and AC have to sum up to A, BA and BB to B and B and A to Total.



- The time series  $\{y_{it}\}_{i \in \{1, \dots, 8\}}$  underlies a linear restriction.
- The restrictions can be comprised into the following equation:

$$y_t = \underbrace{\begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 \\ & & I_5 \end{pmatrix}}_{=:S} b_t, \quad (1)$$

- $b_t = (AA_t, AB_t, \dots, BB_t)^T$  are called *bottom time series*,
- $S$  is the *Summing Matrix*,
- In general:  $y_t \in \mathbb{R}^n, b_t \in \mathbb{R}^m, S \in \{0, 1\}^{n \times m}$ .

- Now we (point-)forecast the time series  $h$ -steps ahead
- We obtain  $\{\hat{y}_{it}(h)\}_{i \in \{1, \dots, n\}}$ .
- In general, these *base forecasts* are *incoherent*, meaning that they do not fulfill the hierarchical restrictions, i.e.

$$\hat{y}_t(h) \neq S\hat{b}_t(h). \quad (2)$$

- Coherency can be enforced by constructing the *reconciled forecasts* as

$$\tilde{y}_t(h) = S\tilde{b}_t(h) = SP\hat{y}_t(h), \quad S \in \mathbb{R}^{n \times m}, P \in \mathbb{R}^{m \times n} \quad (3)$$

- The question is: How to choose  $\tilde{b}_t$ , respectively  $P$ ?

# Reconciliation Strategies

## Reconciliation strategies from Hyndman et al., 2011

- 1) Bottom-up reconciliation:  $\tilde{b}_t(h) = \hat{b}_t(h)$ , respectively  $G = (0_{m \times (n-m)} | I_m)$ .
- 2) Top-down reconciliation:  $G = ((p_1, \dots, p_m)^T | 0_{m \times (n-1)})$ , with  $\sum_{i=1}^m p_i = 1, p_i > 0 \forall i \in \{1, \dots, m\}$ .



### 3) GLS reconciliation:

- Cast the relation between  $\hat{y}_t(h)$  and  $b_t(\tilde{h})$  as regression problem,

$$\hat{y}_t = S\tilde{b}_t(h) + \varepsilon_t \quad (4)$$

- Find  $\tilde{b}_t(h)$  such that MSE is minimized.
- The BLUE for  $\tilde{b}_t(h)$  is the GLS estimator  $(S^T \Sigma^{-1} S)^{-1} S^T \Sigma^{-1} \hat{y}_t(h)$ , where  $\Sigma = \text{Cov}(\varepsilon_t, \varepsilon_t^T)$ .

$$\Rightarrow \tilde{y}_t(h) = S(S^T \Sigma^{-1} S)^{-1} S^T \Sigma^{-1} \hat{y}_t(h). \quad (5)$$

## Unbiasedness

- Let  $\hat{y}_t(h)$  be unbiased predictions,
- So  $\mathbb{E}(\hat{y}_t(h)|\mathcal{F}_t) = \mathbb{E}(y_t(h)|\mathcal{F}_t) = S\mathbb{E}(b_t(h)|\mathcal{F}_t) =: S\beta_t(h) (*)$
- We require that also  $\tilde{y}_t(h) = SP\hat{y}_t(h)$  should be unbiased,

$$\mathbb{E}(\tilde{y}_t(h)|\mathcal{F}_t) = SP\mathbb{E}(\hat{y}_t(h)|\mathcal{F}_t) \stackrel{*}{=} SP S\beta \quad (6)$$

- Now [Wickramasuriya et al., \(2019\)](#) show that  $\Sigma$  is unidentifiable.
- To see this notice that from

$$\tilde{\varepsilon}_h = \hat{y}_{t+h} - \tilde{y}_{t+h} = (I - SG)\hat{y}_{t+h} \quad (7)$$

we get

$$\text{Var}(\varepsilon_h | \mathcal{F}_t) = (I - SG)\Sigma_h(I - SG)^T \quad (8)$$

- $(I - SG)$  is not invertible, this follows from  $SGS = S$ , which is a consequence of requiring unbiasedness.
- GLS reconciliation is not possible!

# MinT Reconciliation

## MinT Reconciliation

Minimum Trace (MinT) reconciliation:

- Minimize  $\text{tr}(\text{Var}(y_{t+h} - \tilde{y}_t(h))) = \text{tr}(\text{Var}(y_{t+h} - SG\hat{y}_t(h)))$  subject to  $SPS = S$ .
- $\Rightarrow \tilde{y}_t(h) = S(S^T W_h^{-1} S)^{-1} S^T W^{-} \hat{y}_t(h)$
- Where  $W$  is the covariance matrix of base forecast errors, not coherency errors.
- The MinT approach is unbiased ( $SPS = S$ ) and
- minimizes the variance of predictions.

- The MinT predictions fulfill:

$$[y_{t+h} - \tilde{y}_t(h)]W_h^{-1}[y_{t+h} - \tilde{y}_t(h)] \quad (9)$$

$$\leq [y_{t+h} - \hat{y}_t(h)]W_h^{-1}[y_{t+h} - \hat{y}_t(h)] \quad (10)$$

- MinT Reconciliation is beneficial beyond coherency.

- The question remains on how to estimate  $W_h$ .
- Possible approaches from the paper ( $k_h > 0$ ):
  - $W_h = k_h I,$
  - $W_h = k_h \text{diag}(\hat{W}_1),$
  - $W_h = k_h, \Lambda = \text{diag}(S\mathbf{1}),$
  - $W_h = k_h \hat{W}_1$

## Open Research

- (Point-)forecast reconciliation by  $\tilde{y}_t(h) = S(g \circ \hat{y}_t(h))$ .
- Probabilistic-forecast reconciliation,
- Temporal Hierarchies,
- Artificial Hierarchies.



## Literature



Hyndman, R. J., Ahmed, R. A., Athanasopoulos, G. & Shang, H. L. (2011)  
Optimal combination forecasts for hierarchical time series  
*Computational statistics & data analysis* 55(9) 2579–2589.



Wickramasuriya, S. L., Athanasopoulos, G., & Hyndman, R. J. (2018).  
Optimal forecast reconciliation for hierarchical and grouped time series  
through trace minimization  
*Journal of the American Statistical Association*.

# Thanks for your Attention!