Multivariate State-Space Approach to Variance Reduction in Series with Level and Variance Breaks due to Sampling Redesigns

THE CASE OF THE DUTCH ROAD TRANSPORTATION SURVEY

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1-4 Sep 2013, SAE, Bangkok

Oksana Bollineni-Balabay, Jan van den Brakel, Franz Palm MULTIVARIATE STATE-SPACE APPROACH TO VARIANCE REDUCT



- Data
- State-Space Structural Modelling Approach
- Competing alternatives
 - Univariate Models
 - A 9-dimensional model
 - A 10-dimensional model
- Signal Variance Comparison
- Conclusions

The subject of study

road freight transportation carried out by vehicles registered in the RDW

- domestic
- own-account
- measured in tons
- quarterly, since 1976
- subdivided into 9 NSTR-domains (Nomenclature uniforme des marchandises pour les Statistiques de Transport, Revise)

NSTR categories

NSTR 0: Agricultural products and live animals;

NSTR 1: Foodstuff and animal fodder;

NSTR 2/3: Solid mineral fuels; Petroleum oils and petroleum;

NSTR 4: Ores, metal scrap, roasted iron pyrites;

NSTR 5: Iron, steel and non-ferrous metals (including intermediates):

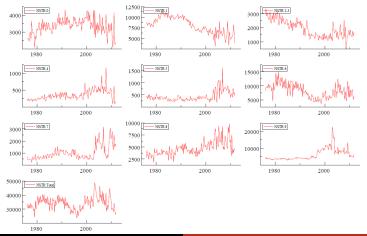
NSTR 6: Crude and manufactured minerals, building materials;

NSTR 7: Fertilizers;

NSTR 8: Chemicals;

NSTR 9: Vehicles, machinery and other goods (including cargo).

Horvitz-Thompson Estimates of the Own Account Transportation Series, 1000 tons



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Major Amendments to the Survey design

- up until 2003: sampling unit in stratified sampling scheme the vehicle;
- 2003-2007: 2-stage stratified sampling design; PSU the company;
- from 2008: back to 1-stage stratified sampling design; sampling unit the vehicle;
- decreasing sample sizes throughout the course of the survey

Modelling Alternatives

- 10 univariate models;
- a 9-dimensional multivariate model ⇒ get the total series by summing up the 9 series estimates;
- a 10-dimensional model: 9 domains and 1 national level series.

Decomposition into Unobserved Components

Horvitz-Thompson estimates:

$$\hat{Y}_{d,t} = \theta_{d,t} + e_{d,t} \tag{1}$$

 θ is the true value of the population variable, e_t is a sampling error.

$$\theta_{t,d} = \underbrace{L_{t,d} + \gamma_{t,d} + \mathbf{x}'_{t,d}\boldsymbol{\beta}_{d}}_{signal} + \underbrace{\varepsilon_{t,d}}_{irregular \ term}$$
(2)
$$\hat{Y}_{t,d} = \underbrace{L_{t,d} + \gamma_{t,d} + \mathbf{x}'_{t,d}\boldsymbol{\beta}_{d}}_{\boldsymbol{\alpha}_{t,d}} + \underbrace{\varepsilon_{t,d} + e_{t,d}}_{\boldsymbol{\nu}_{t,d}}$$
(3)

 $L_{t,d}$ -trend component; $\gamma_{t,d}$ - seasonal component; $\mathbf{x_{t,d}}$ - K (dummy)regressors; $\beta_{\mathbf{d}}$ - K regression coefficients.

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Univariate Models 9-dimensional Model 10-dimensional Model

Univariate Model Estimated

$$\hat{Y}_{t,d} = L_{t,d} + \gamma_{t,d} + x_{t,d}\beta_d + \nu_{t,d}$$

- point-estimates nearly identical to those in multivariate settings;
- variance estimates have a potential for improvement

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Level and Variance breaks

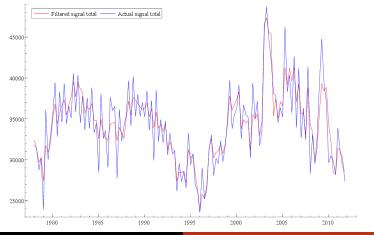
	Level interventions	Number of breaks in $\sigma^2_{\nu,t,d}$		
NSTR 0	-	1		
NSTR 1	-	1		
NSTR 2/3	2008(3)-(4)	2		
NSTR 4	-	1		
NSTR 5	-	2		
NSTR 6	-	2		
NSTR 7	2003(1)-2010(4),	1		
	2007(1)-2008(4)			
NSTR 8	-	1		
NSTR 9	1997(1)-2002(4),	2		
	2003(1)-(4)			
Total	2003(1)-(4)	4		

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Horvitz-Thompson vs. Filtered Signal Estimates of the Total Series from the Ten-Dimensional Model; 1000 tons

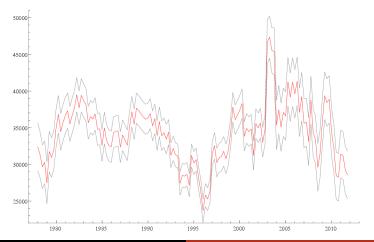


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Filtered Signal Estimates of the Total Series from the Ten-Dimensional Model, 1000 tons

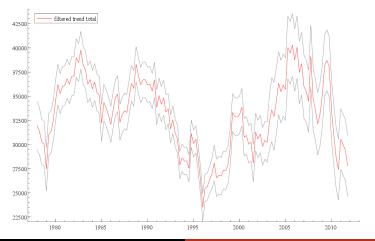


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Filtered Trend Estimates of the Total Series from the Ten-Dimensional Model, 1000 tons

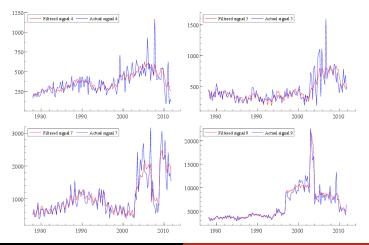


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Filtered Signal Estimates from the Ten-Dimensional Model vs. Horvitz-Thompson Estimates, 1000 tons

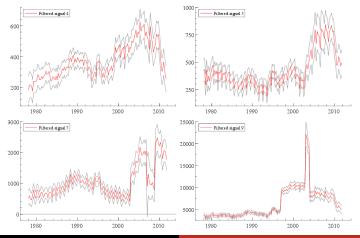


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Filtered Signal Estimates from the Ten-Dimensional Model, 1000 tons

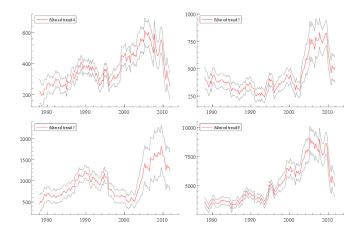


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Filtered Trend Estimates from the Ten-Dimensional Model, 1000 tons

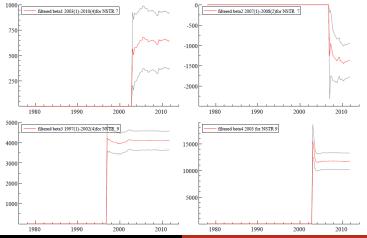


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Filtered level break estimates from the ten-dimensional model (NSTR 7 and 9), 1000 tons



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9-dimensional Model Estimated

$\hat{\mathbf{Y}}_{t} = \mathbf{L}_{t} + \gamma_{t} + \mathbf{x}_{t,1}\beta_{1} + ... + \mathbf{x}_{t,5}\beta_{5} + \nu_{t}$

Cointegration concept implementation (common factor model):

- D trends are driven by p < D stochastic factors;
- dependent trends expressed as a linear combination of the other trends.

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Common Factor Model

Cointegration detection:

modelling covariances between the slope disturbances $\eta_{R,t,d}$, $\eta_{R,t,d'}$ through the Cholesky decomposition

$$\mathbf{Q}_{\mathsf{R}} = \mathsf{E}(\eta_{\mathsf{R}}\eta'_{\mathsf{R}}) = \begin{pmatrix} Q_{11} & Q_{12} & Q_{13} & \cdots & Q_{19} \\ Q_{21} & Q_{22} & Q_{23} & \cdots & Q_{29} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ Q_{91} & Q_{92} & Q_{93} & \cdots & Q_{99} \end{pmatrix} = \mathsf{ADA'}$$

an eigenvalue d_{ii} close to zero \Rightarrow dependent trend

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Common Factor Model

- reveals a relationship between the domains;
- model parsimony;
- variance reduction.

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Common Factor Model

Dependent trends: NSTRs 4, 7, 8 and 9

	(d ₁₁	0	0	0	0	0	0	0	0\
D =	0	d ₂₂	0	0	0	0	0	0	0
	0	0	d ₃₃	0	0	0	0	0	0
	0	0	0	0	0	0	0	0	0
	0	0	0	0	d_{55}	0	0	0	0
	0	0	0	0	0	d ₆₆	0	0	0
	0	0	0	0	0	0	0	0	0
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	0 /	0	0	0	0	0	0	0	0/

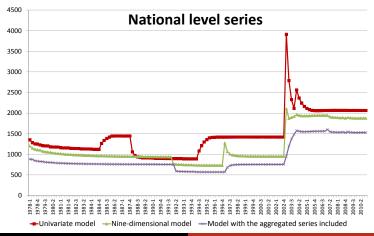
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10-dimensional Model

Why at all?

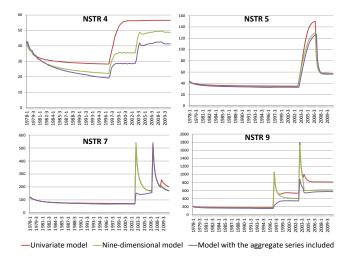
- secures $\sum_{d=1}^{9} \hat{Y}_{d,t} = \hat{Y}_{Total,t}$
- more efficient Kalman filtering \Rightarrow reduced variance in estimates
- Additional complications/assumptions:
 - proper restrictions on the structure of the covariance matrix of disturbance terms;
 - composite error terms:
 - non-constant variances;
 - $\bullet \Rightarrow \mathsf{assumed}: \mathsf{constant} \mathsf{ conditional} \mathsf{ correlation}.$
 - $\Rightarrow\Rightarrow$ for simplicity and with little loss in estimate precision, these covariances are set to zero.

SE of Filtered Signal Estimates; 1000 tons



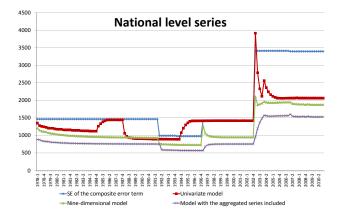
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SE of Filtered Signal Estimates; 1000 tons



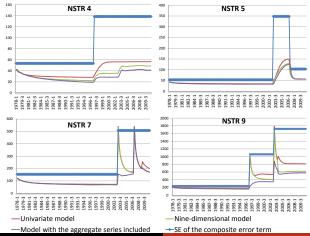
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SE of Filtered Signal Estimates and of Measurement Equation Error Term; 1000 tons



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SE of Filtered Signal Estimates and of Measurement Equation Error Term; 1000 tons



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- Two problems solved simultaneously:
 - breaks (in the level and variance);
 - small sample sizes.
- The signal variance gets reduced when one moves from the univariate models to multivariate ones.

 \Rightarrow The 10-dimensional model with the aggregate series outperforms all the other models.

Thank you!

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