

# Incorporating macro-economic leading indicators in inventory management

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# Motivation

When will the next economic crisis hit? Where? For how long?



Traditional univariate forecasting techniques do not incorporate context information

# Research Question

Long term sales forecasting are formulated using

- Historical data patterns (level, trend, seasonality, ...)
- Promotions
- Judgemental adjustments:
  - Collaborative input from clients
  - Newspapers and industry magazines
  - Rumors in the corridors

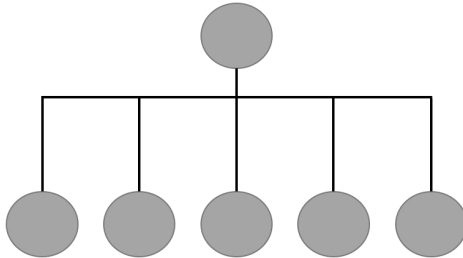
Judgemental input is known to be biased and inconsistent  
(Fildes and Goodwin 2007, Trapero et al. 2013)

- Information of exogenous leading indicators
  - Capturing market sentiment in external big data (Russom et al. 2011)

# Research Question

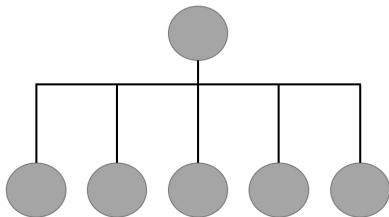
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- Can macro-economic indicators improve sales forecasts?



- What is the real impact on the supply chain inventory?

# Experiment design



Incorporating leading indicator information

- Tactical level
- Plant level
- Top-down level

Evaluation: MAPE and MdAPE

# Models

## Benchmark models

- Naive model
- Holt-Winters model
- Exponential Smoothing

## LASSO model

$$\hat{Y}_i = \beta_0 + \sum_{k=1}^S \beta_k D_k + \sum_{j=1}^P \beta_j x_{ij} \quad (1)$$

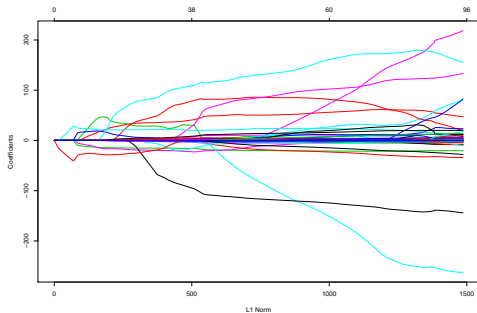
Cost function:

$$\sum_{i=1}^n \left( y_i - \beta_0 - \sum_{p=1}^P \beta_p x_{ip} \right)^2 + \lambda \sum_{p=1}^P |\beta_p| \quad (2)$$

# LASSO

## Least Absolute Shrinkage Selection Operator (Tibshirani, 1996)

- Shrinkage and variable selection
- Selecting  $\lambda$  through cross-validation



## Working paper:

Sagaert Y. R., Aghezzaf E.H., Kourentzes N. and Desmet B.  
Tactical sales forecasting using a very large set of macroeconomic indicators. European Journal of Operational Research.

- MAPE improvement 18.8% on 1-12 months ahead
- Set of 67,851 indicators
- Unconditional Forecasting
- Final model: 10-15 indicators selected
  - Employment in automobile
  - National passenger car registrations
  - Consumer Prices Index for solid fuel prices

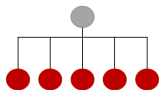


## Sales data of 5 plants of a global manufacturer

- Train period: 2005 - 2012
- Test period: 2013 - 2014
- Forecast horizon  $h=1..6$
- Rolling origin evaluation

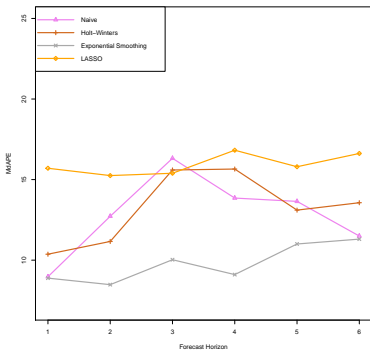
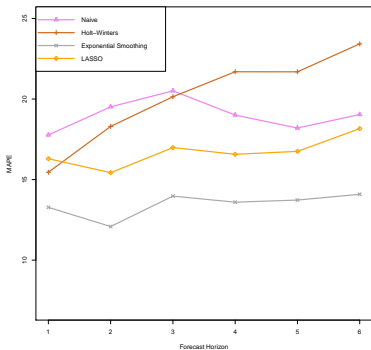
# Empirical results: forecasting accuracy

Lower  
level



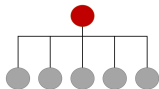
	MAPE	MdAPE
Naive	19.0	12.8
Holt-Winters	20.1	13.2
Exponential smoothing	13.5	9.8
LASSO	16.7	15.9

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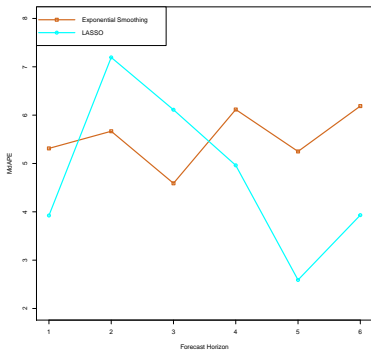
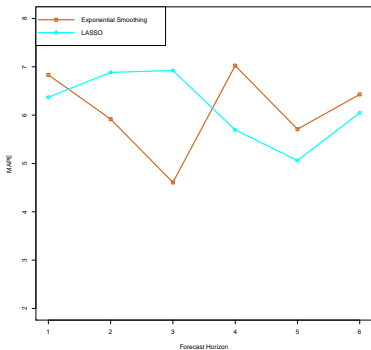
# Empirical results: forecasting accuracy

Higher  
level



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	MAPE	MdAPE
Exponential smoothing	6.1	5.5
LASSO	6.2	4.8

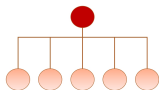


# Reconciliation hierarchical forecasting

The hierarchy is captured in the summing matrix  
Reconciliation incorporates  $1/MSE$  of each forecast

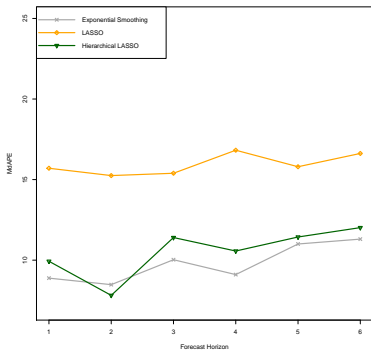
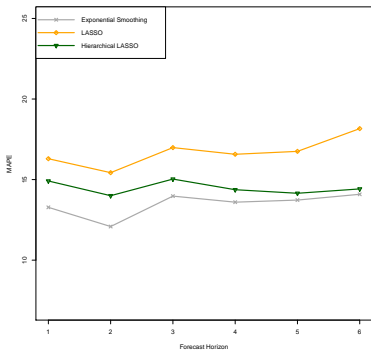
$$\begin{bmatrix} \hat{Y}_{Tot} \\ \hat{Y}_A \\ \hat{Y}_B \\ \hat{Y}_C \\ \hat{Y}_D \\ \hat{Y}_E \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \hat{Y}_{A,r} \\ \hat{Y}_{B,r} \\ \hat{Y}_{C,r} \\ \hat{Y}_{D,r} \\ \hat{Y}_{E,r} \end{bmatrix}, \quad (3)$$

# Empirical results: forecasting accuracy



Reconciled  
lower  
level

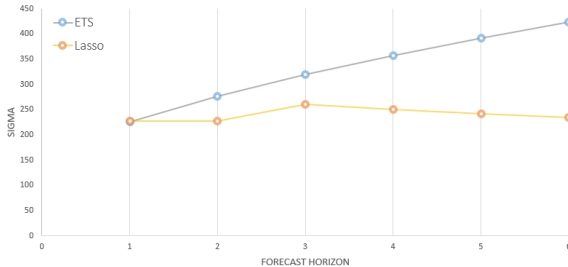
	MAPE	MdAPE
Exponential smoothing	13.5	9.8
LASSO	16.7	15.9
Hierarchical LASSO	14.5	10.5



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# Uncertainty: iterative vs direct forecasting

Reformulated LASSO model for each horizon allows for empirical estimation of  $\sigma_h$

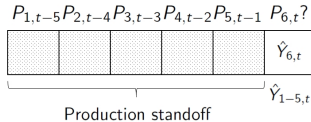


Direct forecasting: independent across horizons

Iterative forecasting: covariances inflate variance

# Inventory simulation

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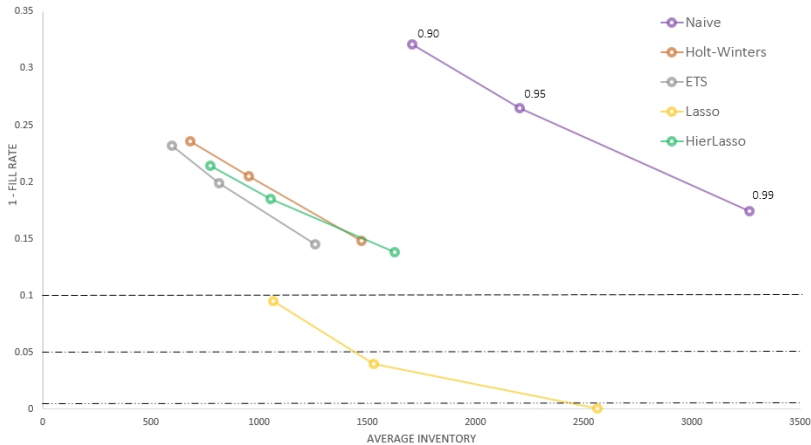


## Simulation parameters

- Production standoff  $t+6$
- Service level: 0.9, 0.95, 0.99
- Inventory policy: Make to stock

# Average inventory per service level

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# Conclusion

- LASSO has an improved forecasting accuracy on long-term
- On short horizons, LASSO leads to service level and inventory improvements

# Questions?

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# Thank you for your attention !

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