

Sparse Warcasting

Forecasting in a data-rich but
statistics-poor environment

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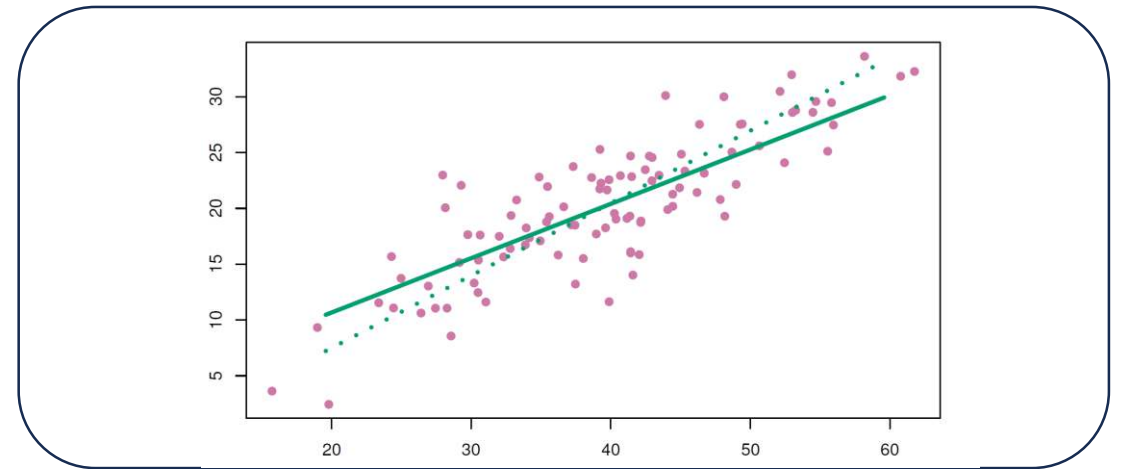
National Bank of Ukraine & UvA

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Traditional Dynamic Factor Models

- Dynamic Factor Models are estimated using **Principal Components**, an unsupervised factor extraction method
- **PCR** factors, a dimensionality reduction technique, is an **unsupervised** method (PC factors - dotted)
- **Forecasting/nowcasting** is a **supervised** application (PLS factors - solid)



$$\begin{aligned}x_t &= \Lambda f_t + \xi_t, & \xi_t &\sim \mathcal{N}(0, \Sigma_\xi) \\f_t &= \sum_{i=1}^p A_i f_{t-i} + \zeta_t, \\ \zeta_t &= B \eta_t, & \eta_t &\sim \mathcal{N}(0, I_q) \\ \hat{y}_t &= \beta' f_t.\end{aligned}$$

Supervised DFM – The Algorithms

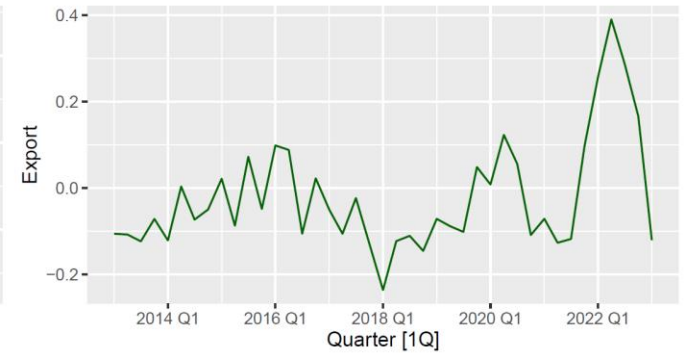
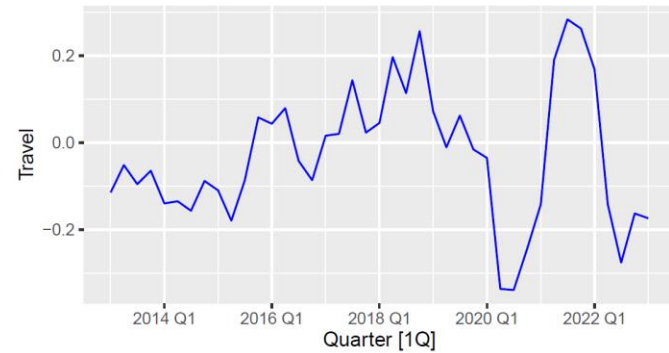
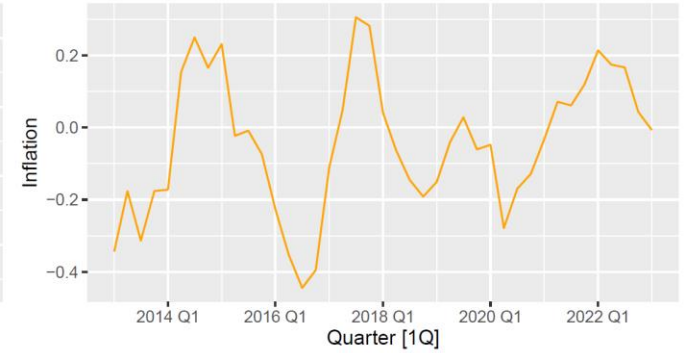
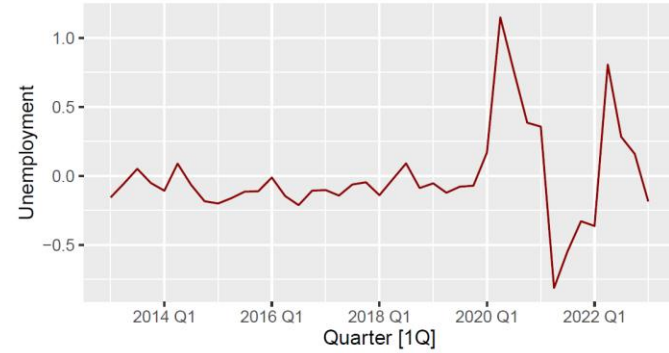
- PCR, extract factors to be independent of each other **regardless** of who \mathbf{y} is.
- PLS, being supervised by \mathbf{y} , identifies the components or factors to be independent of each other while also having high correlation with the target \mathbf{y} {Wold et al. 1984}

$$\begin{aligned} & \max_{\alpha} \text{Var}(\mathbf{X}\alpha) \\ & \text{subject to } \|\alpha\| = 1, \alpha^T \mathbf{S}v_{\ell} = 0, \ell = 1, \dots, m-1, \end{aligned}$$

$$\begin{aligned} & \max_{\alpha} \text{Corr}^2(\mathbf{y}, \mathbf{X}\alpha) \text{Var}(\mathbf{X}\alpha) \\ & \text{subject to } \|\alpha\| = 1, \alpha^T \mathbf{S}\hat{\varphi}_{\ell} = 0, \ell = 1, \dots, m-1. \end{aligned}$$

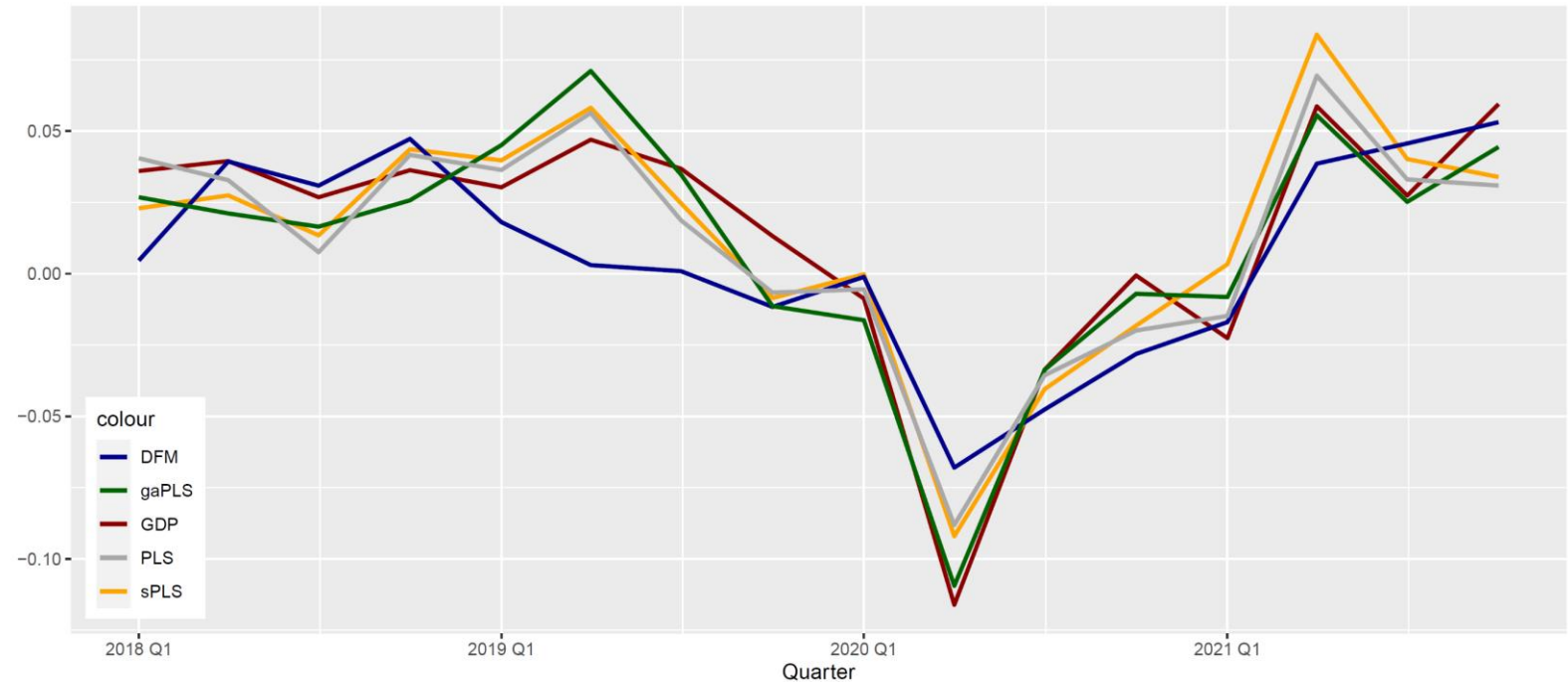
Data

- **Input:** A list of Google Trends time-series: Topics and Keywords
- **Target:** quarterly deflated y-o-y GDP from 2012 to 2021
- **Objective:** Nowcast and near-forecast GDP in the early stages of the Russian 2022 invasion

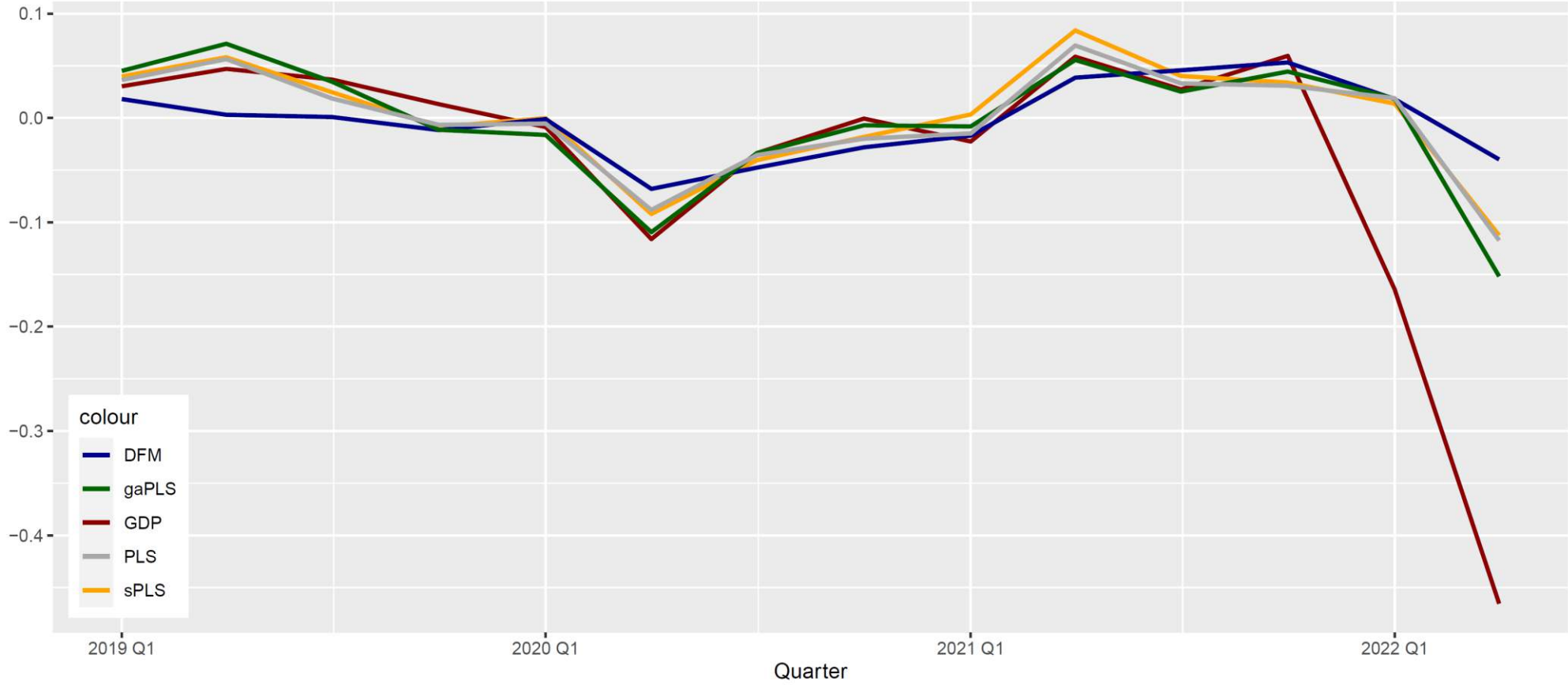


Methodology & In-sample Fit

- Variable selection is performed by a Genetic Algorithm, several alternative also considered
 - Sparse PLS by Chun & Keles and Filter methods
- The algorithm considered **the joint problem** of finding relevant variables & the appropriate number of latent factors



Out-of-sample



Selected Variables



Conclusions

- Preliminary results show the value of using supervised algorithms in conjunction with variable selection methods to nowcast and near-term forecast when no official statistics are available
- PLS outperforms PCR primarily thanks to its better identification of latent factors
 - Lower number of factors also relevant given the small sample