Using the payment system data to forecast the Italian GDP

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Outline

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2 Introduction



Forecasting application
The Model
Results



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Search for innovative data sets to improve the forecasting ability Following the financial crisis, increasing need to link financial information with macro forecasts

Data from Payment system show suitable features to track the short-term evolution of the economic activity

Compiled in real-time by a single source \rightarrow timeliness and reliability No backward revisions

Relevant amount compared to the GDP • Figure

Describes the payment system data. A picture for Italy

Provides empirical evidence on the comovement between payment data and economic activity

Evaluates the ability of the payment data to forecast GDP q-o-q

- selection of targeted predictors LASSO
- MIDAS dynamic factor model with Kalman smoothing

Proves the contribution of the payment data to track the short-term evolution of the $\ensuremath{\mathsf{GDP}}$

Duarte, Rodrigues e Rua 2016

ATM/POS data for nowcasting and forecasting the quarterly private consumption. MIDAS model.

Galbraith and Tkacz 2013

Values and volumes of debit and credit card as well as cheques to nowcast GDP growth for Canada. Simple OLS model with three regressors (Composite Leading Index, PD, lagged y).

Carlsen and Storgaard 2010 *Electronic payments by card to nowcast the retail sales in Denmark.*

Esteves 2009 ATM/POS data as an indicator to estimate private consumption (the y-o-y growth of non-durable goods and consumption) in Portugal. Single-indicator models and different forecast combinations (Diebold 1988) assessed.

Galbraith and Tkacz 2009 *Canadian debit card data as high-frequency indicators to analyze the very short-term impact of quite large and highly transitory economic events (extreme events as the terrorist attacks on 9/11 and electrical blackout on 8/13) on the economy. Only descriptive evidence.*

Galbraith and Tkacz 2007 *Debit cards payments to monitor the economic activity in real-time (correlation between PD and a) consensus forecasts errors, b) revisions on the first official estimate; the short-term impact of extreme events. No inference; only empirical evidence.*

비 수준에 준답

The set of instruments, procedures, rules and intermediaries that enable the circulation of money

Payments classified according to agents making the payment (P. Schioppa, Seoul May 2004)

Wholesale $\,\rightarrow$ within the banking system

Retail \rightarrow within the circuit of **individuals and firms**

Different needs depending on the size and on the urgency of the payments

Monetary union entailed the harmonization of the infrastructure to transfer money

effectiveness of the monetary policy

security of the payments

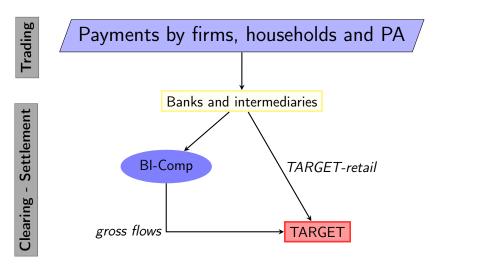
Wholesale payments

1999 **TARGET** 2003 **TARGET2** (crucial contribution of the Bol)

Retail payments

multiple inter-countries links **BI-Comp** pan-European platform **STEP2** 2014 **SEPA**

Payment system - retail payment value chain



Cash payments

Non-cash payments

Distinction based on the party submitting the payment order

- credit-based submitted by the payer
- debit-based submitted by the payee

Cash-payment is still highly widespread in terms of volume: 80% of the number of transactions

- $\ \ldots$ but it accounts for 45% of the value of consumer-to-business transactions
- and 10% of all transactions (including business-to-business)

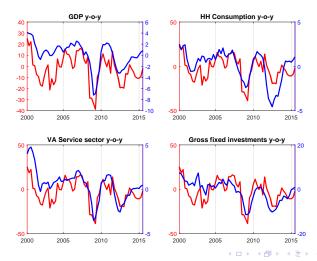
Retail non-cash payments settled through BI-Comp and T2 amount to 5 trillion \notin on yearly basis \bullet Figure

- 60% of the total value of retail non-cash payments
- 80% if we consider only the electronic payments

Credit transfer represents the largest share of the total non-cash transactions (55%)

Payment data and macroeconomic aggregates

Strong empirical evidence on the tight relationship between macro aggregates (bl) and payment flows (rl)



Payment data and macroeconomic aggregates

Strong empirical evidence on the tight relationship between macro aggregates (bl) and payment flows (rl)

Payment flows	GDP	Private Consumption	Gross Fixed Investment	Value added service sector	
Cheques	-44.4	-2.1	-37.2	-35.6	
Cards	1.4	-12.3	16.4	4.0	
Credit Transfer	59.1	61	65.3	59.5	
Direct Debts	52.1	39.4	48.7	54.4	
BI-Comp	71.8	70.5	74.3	68.4	
TARGET-retail	79.4	67.7	64	74.3	
BI-Comp + T2	90.7	82.5	82.3	85.9	
Other indicators					
M2	-38.1	-11.5	-20.4	-45.6	
Industrial production	92.8	68.7	74.3	79.8	
Business Confidence	68.5	74.9	67.3	64.3	
Consumer Confidence	9.9	40.0	14.3	19.7	

Correlation between payment flows and macroeconomic indicators

We started to compile N = 50 variables LASSO selected n_L targeted predictors

$$\hat{L}_{n_L} = \{ j \in \{1, 2, \dots, N\} : |\hat{\beta}_{L_j}| > 0 \}$$

Post-LASSO selection $\hat{I} \supseteq \hat{L}_{n_L}$

▶ Data

Factor-MIDAS with Kalman smoothing

The dynamic factor model

$$\mathsf{X}_t = \Lambda_s(L)\mathsf{f}_t + \xi_t$$

can be mapped into the...

... static factor model

$$\mathbf{X}_t = \mathbf{DF}_t + \xi_t$$

where

$$\mathbf{F}_t = \mu + \Psi_1 \mathbf{F}_{t-1} + \dots + \Psi_l \mathbf{F}_{t-l} + \mathbf{u}_t$$

We assume

$$\mathbf{u}_t = \mathbf{R}\epsilon_t$$

 \mathbf{u}_t has dimension r and ϵ_t is a q-vector of primitive shocks. Σ_u has rank q < rUnrestricted MIDAS model

$$y_{t_q} = c + \beta_0 \mathbf{F}_{t_q} + \beta_1 \mathbf{F}_{t_q-1/m} + \dots + \beta_p \mathbf{F}_{t_q-p/m} + \epsilon_{t_q}$$

Forecasting application

We compare \hat{l} and \hat{l}_{-PS} out-of-sample forecasts (pseudo real-time simulations) Rolling window (length 78 months)

	RMSFE								
Sample [2008.Q2 - 2015.Q2]									
	Backcast	Nowcast	Forecast 1-step						
î	0.909	0.984	0.991						
\hat{I}_{-PS}	0.911	1.028	1.038						
	Sample [2011.Q3 - 2015.Q2]								
î	0.517	0.529	0.544						
Î_PS	0.528	0.543	0.572						

RMSFE of the models \hat{I} and \hat{I}_{-PS}

The smoothed state vector can be expressed as the weighted sum

$$\mathsf{FF}_{t|T} = \sum_{j=1}^{T} w_j (\mathsf{FF}_{t|T}) \mathsf{X}_j$$

Let us assume current month November 2015 backcast GDP growth on 2015.Q3 nowcast 2015.Q4 one-step-ahead forecast 2016.Q1



The flows of retail payments

track the short-term evolution of the economic activity stand out among other business cycle indicators

Caveat

the structure of the retail payment system is changing after the introduction of the SEPA standards

Way forward

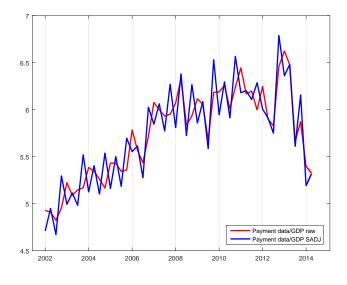
density forecast assessment

forecast of both the supply/demand-side components of the GDP

THANK YOU

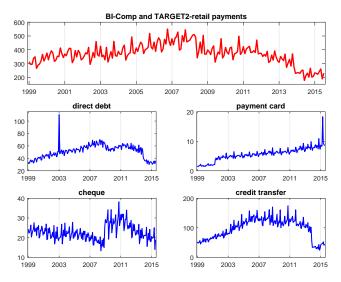


Ratio between Payment flows and GDP



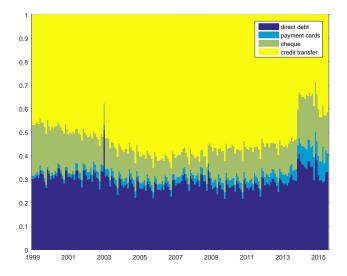
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Total payments settled in BI-Comp and TARGET2-retail (trillion \in)



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Quotas of the payment instruments



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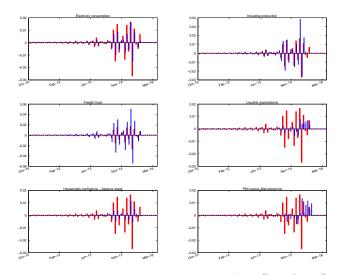
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N-sample compared to LASSO and post-LASSO samples

Indicators	Source	Treatment	Lasso selected	Added to Lasso selection
Total electricity consumption	Italian Electrical Net Society	(1-L)logSA	yes	yes
Industrial production	Istat	(1-L)log	yes	yes
Business Climate	Istat	SA	yes	yes
CCS - Future General Economic sit.	Istat	SA	yes	yes
CCS - Future Personal Economic sit.	Istat	SA	yes	yes
CCS - Unemployment exp.	Istat	SA	no	yes
CCS - Saving opportunities, next 12 mth	Istat	SA	yes	yes
CCS - Households balance sheet	Istat	(1-L)	yes	yes
CCS - Current saving opportunities	Istat	SA	no	yes
BCS - Current level of orders (intermediate goods)	Istat	SA	no	yes
BCS - Current level of production (int. goods)	Istat	SA	no	yes
BCS - Exp. level of production (int. goods)	Istat	SA	no	yes
BCS - Exp. level of orders (int. goods)	Istat	SA	no	yes
BCS - Future general economic sit. (int. goods)	Istat	SA	no	yes
BCS - Exp. level of liquidity (int. goods)	Istat	SA	no	yes
BCS - Current level of liquidity (int. goods)	Istat	SA	no	yes
BCS - Current level of orders (investment goods)	Istat	SA	no	yes
BCS - Current level of production (inv. goods)	Istat	SA	no	no
BCS - Current level of liquidity (inv. goods)	Istat	SA	no	no
BCS - Exp. level of orders (inv. goods)	Istat	SA	no	no
BCS - Exp. level of production (inv. goods)	Istat	SA	no	no
BCS - Future General Economic sit. (inv. goods)	Istat	SA	no	no
BCS - Exp. level of liquidity (inv. goods)	Istat	SA	no	no
BCS - Current level of orders (consumer goods)	Istat	SA	no	no
BCS - Current level of production (cons. goods)	Istat	SA	no	no
BCS - Exp. level of orders (cons. goods)	Istat	SA	ves	VES
BCS - Exp. level of production (cons. goods)	Istat	SA	no	no
BCS - Future General Economic sit. (cons. goods)	Istat	SA	ves	yes
BCS - Exp. level of liquidity (cons. goods)	Istat	SA	no	
BCS - Current level of liquidity (cons. goods)	Istat	SA		
Current accounts deposits (stock)	Bank of Italy	(1-L)logSA	no	no
Credit flows to firms	Bank of Italy	(1-L)logSA	no	no
HICP	Istat	(1-L)logSA	00	yes
FTSE Italy (Banks)	Datastream	(1-L)log		
FTSE Italy (Insurance)	Datastream	(1-L)log	no	
FTSE Italy (Transport)	Datastream	(1-L)log		
PMI Services - Business activity	Markit	SA	no	no
PMI Services - New business	Markit	SA	yes	yes
PMI Manufacturing - New orders	Markit	SA	ves	yes
PMI Manufacturing - Output	Markit	SA	no	
PMI Manufacturing - New orders	Markit	SA	00	10
PMI Manufacturing - Employment	Markit	SA	no	no
PMI Manufacturing - Employment PMI Manufacturing - New export orders	Markit	SA	no	no
Freight truck	ASPI	(1-L)logSA	yes	yes
Retail trade - goods	Confrommercio	SA SA	yes	10
Retail trade - goods Retail trade - services	Confcommercio	SA	no	yes no
BLCOMP	Bank of Italy	(1-L)logSA	no	no
TARGET RETAIL	Bank of Italy	(1-L)logSA	ves	Ves
Payment system-total	Bank of Italy	(1-L)logSA (1-L)logSA	yes	yes
r ayment system-total	Dalik of Italy	(1-L/IOSSA	110	no



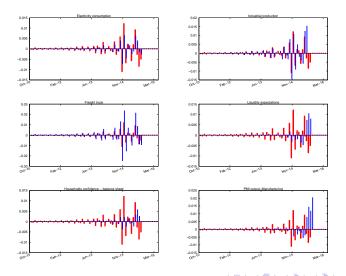
Weights for **backcasting** (PS red bar; other variable blue bar)



Aprigliano, Ardizzi, Monteforte Payments data to forecast GDP

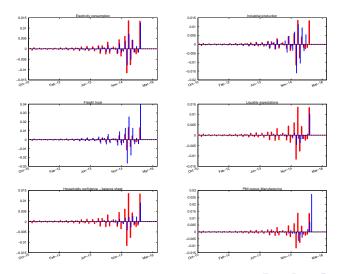
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Weights for **nowcasting** (PS red bar; other variable blue bar)



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Weights for 1-step ahead forecasting (PS red bar; other variable blue bar)



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