

# Using the payment system data to forecast the Italian GDP

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# Outline

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- 1 Motivation
- 2 Introduction
- 3 Dataset
- 4 Forecasting application
  - The Model
  - Results
- 5 Wrap-up

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 → timeliness and reliability

No backward revisions

Relevant amount compared to the GDP [▶ Figure](#)

# This paper

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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 GDP

# Overview of the literature

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## **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.*

# Overview of the literature

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**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.*

# Payment system

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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** → within the **banking system**

**Retail** → within the circuit of **individuals and firms**

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

# Payment system - historical notes

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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

Trading

Payments by firms, households and PA

Banks and intermediaries

BI-Comp

*TARGET-retail*

*gross flows*

TARGET

Clearing - Settlement

# Payment instruments

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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

# A picture for Italy

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

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

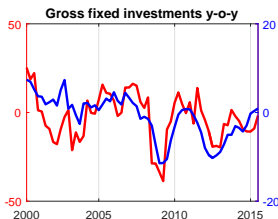
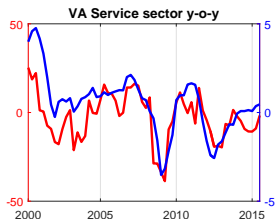
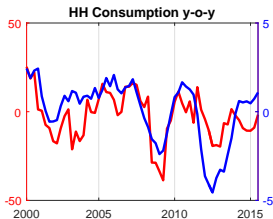
**Retail non-cash payments** settled through BI-Comp and T2 amount to 5 trillion € on yearly basis [▶ Figure](#)

- 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%) [▶ Figure](#)

# 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)

## Correlation between payment flows and macroeconomic indicators

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

# Targeted Predictors

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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**

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

can be mapped into the...

... **static factor model**

$$\mathbf{X}_t = \mathbf{D}\mathbf{F}_t + \xi_t$$

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  $\epsilon_t$  is a  $q$ -vector of primitive shocks.

$\Sigma_u$  has rank  $q < r$

**Unrestricted MIDAS model**

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

# Forecasting application

We compare  $\hat{I}$  and  $\hat{I}_{PS}$

out-of-sample forecasts (pseudo real-time simulations)

Rolling window (length 78 months)

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

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



# Forecasting application

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The smoothed state vector can be expressed as the weighted sum

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

Let us assume current month November 2015

backcast GDP growth on 2015.Q3

nowcast 2015.Q4

one-step-ahead forecast 2016.Q1

▸ Figures

## The flows of retail payments

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

## Caveat

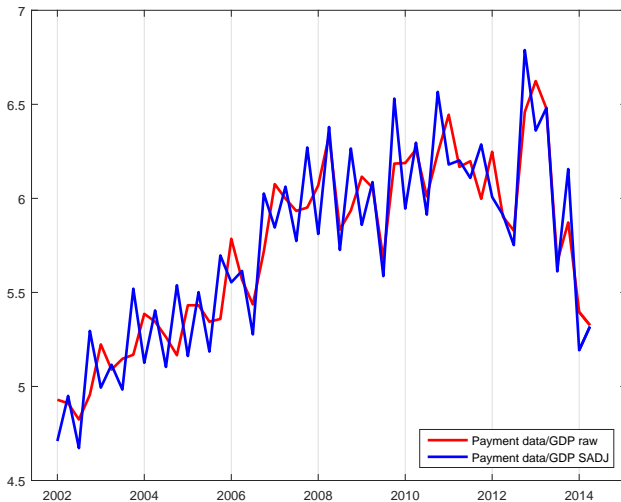
the structure of the retail payment system is changing after the introduction of the SEPA standards

## Way forward

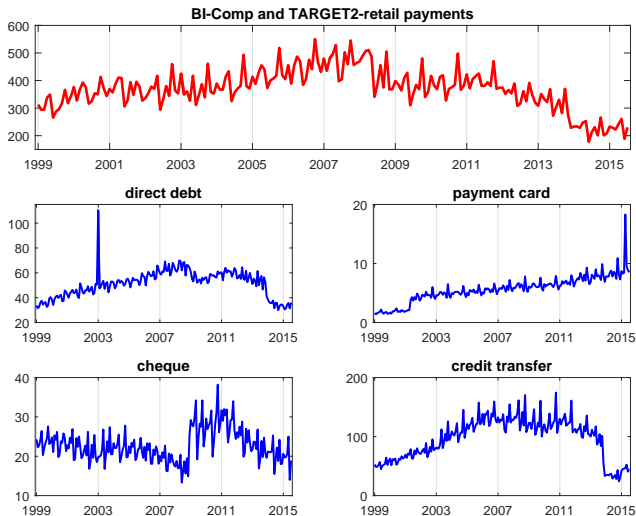
- density forecast assessment
- forecast of both the supply/demand-side components of the GDP

THANK YOU

## Ratio between Payment flows and GDP



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

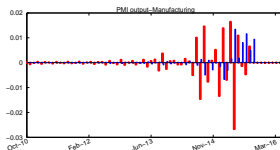
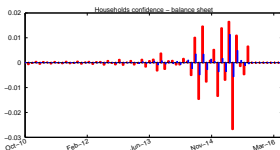
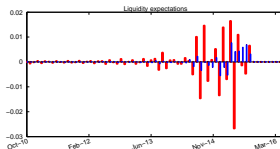
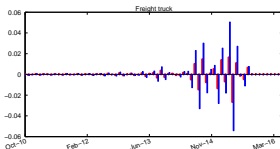
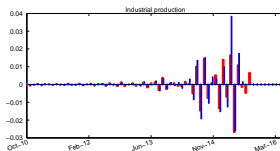
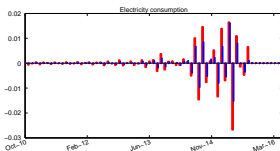


## Quotas of the payment instruments



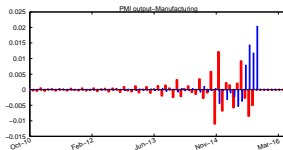
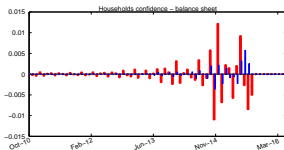
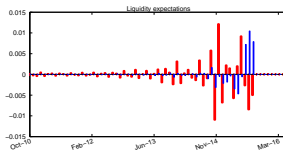
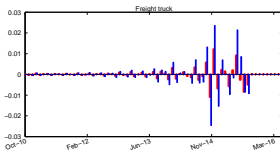
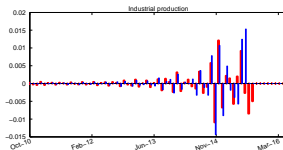
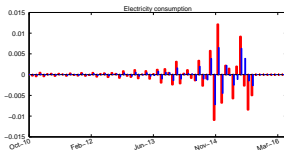


## Weights for **backcasting** (PS red bar; other variable blue bar)





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



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

