National Accounts in a World of Naturally Occurring Data: A Proof of Concept for Consumption

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Advanced analytics: new methods and applications for macroeconomic policy 2022 Bank of England (BoE) - European Central Bank (ECB) – King's College London (DAFM)

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Summary

- Introduction
- Building a Consumption Survey
- National Accounts
- Distributional Accounts
- Potential: Economics of Real Time, High Definition & Smart Policies
- Conclusion

Introduction: Why Use Transaction Data

• Modern payment systems generate an enormous amount of real-time data on activity which can be incorporated into national accounting measures (Bean 2016).

- Advantages of such data include **timeliness, granularity, and cost** to statistical agencies (albeit not to private sector).
- This is **important for lower-income countries**, where transaction data may be the only reliable source of information for building national accounts.
- Growing interest in non-traditional data for tracking the economy, especially in the wake of COVID-19.
- But few, if any, attempts in the academic literature to build national accounting objects from first principles using large-scale payment data.

Introduction: First Proof of Concept on Naturally Ocurring Data

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First proof of concept that naturally occurring transaction data, arising through the decentralized activity of millions of economic agents, can be organized via national accounting rules and then harnessed to produce a large-scale, highquality and highly-detailed consumption survey

- Universe of BBVA retail accounts in Spain by BBVA
- Allowing us to track expenditure as it flows out of these accounts, transaction by transaction
- 3 billion individual transactions by 1.8 million BBVA customers, from 2016 to 2021



Contributions: Four contributions of the Paper

1. How to construct representative panel of household expenditure. Massive survey.

- 2. Show that it aggregates to Quarterly National Accounts
- 3. Create Distributional National Accounts for Consumption

4. Study the Micro-Structure of consumption dynamics (just an example !!):

Building a Consumption Survey

Building a Consumption Survey: Key Problems

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- Two problems for translating transaction data into a representation of the population:
 - The client pool of a **Bank** \neq **Population**: **Biases**
 - Spending is not the same than consumption: **Spending** ≠ **Consumption**

Building a Consumption Survey: Active Customers

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Sample Frame

• 10,270,041 unique customers (2015-2021)

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- Most spend infrequently or for short periods only.
- Define "Active Customers" as making at least 10 consumption related transactions in each quarter.
- 1,827,866

Customers with transactions and sampling frame of active customers

7.000.000 -6.000.000 -5,000,000 -4,000,000 -3,000,000 -2,000,000 -1.000.000 б Q3 Q1 Q3 Q1 Q3 QЗ Q3 Q1 Q1 Q1 Q1 Q3 2016 2015 2016 2017 2017 2018 2019 2019 2019 2020 2020 2021 quarters Customers with at least a transaction in guarter — Sampling frame of Active Customers

Building a Consumption Survey: Non-Housing Expenditures

Classification of Non-Housing Consumption Spending

- If a transaction is explicitly categorized in one of the 12 COICOPS.
- Follow national accounting principles wherever possible

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Spending Category	Volume of Transactions	Number of Transactions
Offline Card Transactions	60,319 million	1,772 million
Online Card Transactions	11,858 million	313 million
Direct Debits	66,036 million	752 million
Cash Withdrawal	64,592 million	359 million
Transfers excl. rent	11,148 million	15 million



Building a Consumption Survey: Housing Consumption (Rents)

Determination of Housing Consumption Spending

- We locate payment of rental for housing services.
 - Reading of free-text field in direct debits and transfers.
 - Minimum 100 EUR
 - Exclude parking, etc.
 - Payments made in 70 months.
 - 32,127 households.
- Use household covariates to predict monthly rent
 - Income (from BBVA table, six month average)
 - Utility Payments (direct debits)
 - Geography: 327 regions (consolidating postal codes)

Out-of-sample behavior is reasonable with households that are 50-70 months in data.

Variable	Model	Test set
Spending on House Utilities	0.0884	
	(0.0008)	
Income	0.0362	
	(0.0011)	
N of Contract Groups	16,977	15,512
N of Observations	1,134,735	15,512
R ²	0.3911	
Adjusted R ²	0.3765	
Within R ²	0.1200	
Root MSE	204.6144	221.64

Use covariates to IMPUTE housing consumption for the rest of the househods (the vast majority)

Building a Consumption Survey: Estimating Households

• Link clients into perceived household groups.

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- Individuals with whom they share a contract and live in same postal code.
- 1,589,280 household groups

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(Customers with transactions and sampling frame of active customers)



Cosigner Groups with Active Customers in Naturally Occurring Data

	HBS 2016	HBS 2017	HBS 2018	HBS 2019	HBS 2020	BBVA Sample
Households	22,011	22,043	21,395	20,817	19,170	1,589,280
Adults	47,420	47,055	45,328	43,988	40,285	1,827,866

Building a Consumption Survey: Demographics of Active Clients

Demographics of Active Customers

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Research

$$c_{g,a,r} = c_{g,a,r}^{\mathrm{hh}} \left(rac{x_{g,a,r}^{\mathrm{INE}}}{x_{g,a,r}^{\mathrm{BBVA}}}
ight)$$





Building a Consumption Survey: Weighting & Sampling

Weighting & Sampling

- We observe Spending not Consumption within household
- Assume equal spending among active clients within Households, and half the weight of non Active Clients
 - Define cells Gender, Age and Region (5 geographical units per province)
 - Adjust Demographic Weights of cells to Make them Representative

$$c_{g,a,r} = c_{g,a,r}^{\mathrm{hh}} \left(rac{x_{g,a,r}^{\mathrm{INE}}}{x_{g,a,r}^{\mathrm{BBVA}}}
ight)$$

- In occasion we need to create national simple
- We draw $x_{g,a,r}^{INE}$ times from the pool of active client IDs within cell
- Sampling with Replacement

Building a Consumption Survey: Payment-Consumption Connection

Connecting Payments with Consumption

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COICOP Shares by Payment Method

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Aggregating To National Accounts

National Accounts: Similar Levels

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Total Consumption Private: Naturally Ocurring vs Official (level Mill Euros)

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Total Private Consumption: Naturally Ocurring vs Of (QoQ Sadj Growth Rate)





National Accounts: COICOP Distributions in HBS and N. Accounts

Distribution of Spending across COICOP Categories



(a) BBVA vs HBS



(b) BBVA vs National Accounts



National Accounts: Richer Dynamics (Daily) by Payment & Category

Proportions of Consumption by Payment Method (Cumulative 3M, % of Total)



Proportions of Consumption by COICOP (Cumulative 3M, % of Total)



Creating Distributional National Accounts For Consumption

Distribut. Accounts: Consumption's Macro-Consistent distribution

Yearly adult consumption by levels of consumption in 2017; (Weighted sampling procedure, 2017 Euro by percentiles of consumption)



• Macro-consistent, distribution of Consumption.

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- It aggregates into NA
- Follows Picketty et al 2018.



Distributional Accounts: Differences Consumption and Income

Consumption Distribution per Spanish adult in 201 BBVA vs. WID 2017 post-tax income distribution.

Lorenz Curves of BBVA 2017 consumption and WID 2017 post-tax income in Spain.



Distributional Accounts: Right Tail Difference (in Level)

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Comparison Distribution of Consumption with Household Business Survey (HBS)

• Remember: difference in level.

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 Right tail difference: Bigger share of consumption among the people who consumes most.

Distribution of Consumption per adult in BBVA vs. Spanish Household Budget Survey consumption per adult distribution 2017

Distributional Accounts: Different Categories by Consumers

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Distribution across COICOPs and People

Consumption distribution disaggregated by COICOP consumption categories

Distributional Accounts: Time Frequency Matters

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Lorenz Curves of the distribution of consumption across different time frequencies.

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- Consumption is a flow.
- Unite of Time aggregation matters.
- Lumpy.

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• Critical in survey design.

Frequency	Gini index
Daily avg. (2017)*	0.629
Weekly avg. (2017)	0.439
Monthly avg. (2017)	0.338
Quarterly avg. (2017)	0.307
Yearly (2017)	0.281
Pre-Covid 3 Years (2017-2019)	0.273
All 5 Years (2017-2021)	0.265
*30 days sampled randomly	

Distributional Accounts: Consumption Growth Distribution

Distribution of Consumption Growth

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- Consumption Growth Inequality not so apparent in normal years (2017-2019)
- But increasingly apparent during Covid (2019-2020)...

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..and Post-Covid Recovery (2020-2021)

Distributional Accounts: Consumption Growth Dist. by Categories

Relying only on Cards can lead to biases towards Consumption inequality

(only increase in a couple of COICOPS in Normal years)

Growth Rate of Consumption per Percentiles by COICOP Consumption Group

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Individual Consumption Dynamics: Panel Structure Individual Level

Distribution of Consumption Growth

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• Panel Structure at individual level.

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- Very different from Cross-Sections
- In one year (pre-COVID), massive mean reversion
- 2020, 2021, much flatter... becase of rapid reversion during first year.

Potential Uses:

The Economy in Real Time, High Definition and Design of Smart Policies

Cards payments (PoS & OnLine) proved to be very useful to track activity

Source: BBVA Research & Buda et Al (2022). National Accounts in a World of Naturally Occurring Data: A Proof of Concept for Consumption. Forthcoming

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But this view is partial (38%) and biased (upwards) after the COVID..

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Cards can be misleading... Leading to an optimistic view

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Feb-20

Official Retail Trade Index

Ma y-20 Aug-20 Nov-20 Feb-21

Vla y-21

BBVA Big Data (Only Cards)

Feb-22

Vla y-22

Aug-21

Nov-21

... And this bias is transmitted to the fit with official Growth rates

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Spain: Real Private Household Consumption Spain: Real Retail Trade Index (Seasonally Adjusted YoY Growth Rate CPI Deflated)

Ma y-18 Aug-18 Nov-18 Feb-19 Ma y-19 Aug-19 Nov-19

Feb-20

Ma y-20 Aug-20 Nov-20 Feb-21 May-21 Aug-21 Nov-21 Feb-22

Vay-22

But fortunately ... we can correct with Naturally Ocurring Data

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... not only in monthly and quarterly but also in Real Time

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Source: BBVA Research & Buda et Al (2022). National Accounts in a World of Naturally Occurring Data: A Proof of Concept for Consumption. Forthcoming

At Geographical Level: National, Regional, Provincial, Urban and Zip Code...

Source: BBVA Research & Buda et Al (2022). National Accounts in a World of Naturally Occurring Data: A Proof of Concept for Consumption. Forthcoming

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... and "High Definition" (i.e zip code level by Activity, Age, Gender ...)

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...Naturally Ocurring Data are also relevant for the design of SmartPolicies

Spain: Proportion of Consumption per Age Groups (2017

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Spain: Levels of Consumption per person by Age Groups

... as High Granularity can give us a "detailed" vision of reality

Spain: Consumption growth by Income Percentile

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...Just a partial recovery for the Lower income people

...A sharp adjustment in Consumption Growth for "High Income" and the lower ones ...

...Remember they are growth rates.. so Consumption Level recovery Incomplete at the end of 2021

Conclusions

Conclusions

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- The vast amount of data naturally occurring within financial institutions can be harnessed to produce high quality consumption survey.
- Unlike standard consumption surveys, a simple aggregation of the survey's data results in National Accounts levels
- But with arbitrary frequency, and incredibly more dense coverage.
- Not only the survey micro data generates distributional accounts for consumption... it allows an individual panel structure that allows for careful study of consumption dynamics
- Of course, it allows for using covariates (income, ...) to understand the determination of consumption at micro and macro levels.

We aim to use similar procedures to determine Investment, I-O tables, external sector, internal flows, etc...

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