

Evaluation of the impact of mixed mode design on the quality of the estimates of the survey "Aspects of Daily Life"

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- Showing the framework for the analyses carried out to assess the impact of Mixed Mode on Aspect of Daily Life household survey, switched from single to mixed mode
 - Experimental setting (parallel samples single/mixed mode)
 - Analysis of the reasons that explain significant differences in the estimates obtained with the two designs
 - Focus on the analysis of impact of mixed mode on univariate distributions and multivariate structure of the data



Mixed mode in ISTAT social surveys

Mixed mode is the combined use of different data collection techniques in one survey,

- MM is spreading especially in social surveys, to contrast declining response rates and coverage, reducing also the total cost of the surveys
- The use of different data collection techniques helps in contacting different types of respondents in the most suitable way for each of them, so allowing a gain in population coverage and response rate
- ✓ It is not a new way of data collection for ISTAT surveys but its treatment has been faced recently

In ISTAT several situations have occurred so far

- Mixed mode used primarily to address coverage issues of previously single mode CATI surveys (-> web/CATI)
- Mixed mode in longitudinal household surveys to reduce cost and burden (CAPI/CATI) Eu-silc and LFS
- Mixed mode used primarily to reduce survey cost whereas expanding population coverage, through the introduction of web technique in traditionally PAPI surveys
 - "Multipurpose survey on households: Citizens and leisure time" 2015", web/PAPI
 - "Multipurpose survey on households: Aspects of daily life 2017": sequential web/PAPI with a control single mode sample PAPI



Which drawbacks has this choice?

The difficulty of controlling over **mode effect** and the **confounding** between **selection** and **measurement** effects (especially in sequential designs) (De Leeuw, 2005)

- Mode effect refers strictly to measurement error differences due to the mode of survey administration (error of observation)
- A **selection effect** occurs due to the differences in the distributions of the respondents to the alternative modes, (error of observation, desirable aspect of MM strategy)

How and when dealing with mode effect?

- □ Mainly in the planning of the survey (questionnaire and survey design) to limit measurement error
- In the estimation phase
 - to evaluate the accuracy of the estimates over time the estimates must be consistent and comparable with the analogue ones obtained in the previous survey editions, for ensuring that any changes in the time series are exclusively due to real changes of the observed phenomenon
 - to adjust mainly the selection effect, while estimating the measurement effect



Theoretical framework

- □ From an inferential point of view the **selection and measurement effects** need to be investigated separately, to obtain a correct formulation of the total non-sampling error and to apply methods to adjust the estimates of the parameters of interest for the bias effects
- □ The problem of the **confounding** between the two effects is the central theme of the theory of **causal inference** (*Pearl, 2009*)
 - The measurement error is conceptualized as a causal effect of the mode on the survey variable, while the selection effect is seen as a spurious correlation between the target variable and the mode
 - For the estimation of the two effects causal inference is used according to a counterfactual perspective: the existence of a potential result not really observed (the value that the respondent would have provided with the other mode) is hypothesized
- An alternative approach is based on the use of **instrumental variables**, when a benchmark survey is available (*Vannieuwenhuyze et al. 2010*)



The sample survey "Multipurpose survey on households: Aspects of daily life"

- Collects information about recreational and cultural activities in free time, such as sports, reading, cinema, music, the Internet, social relations, issues for the quality of people life
- Based on <u>a sample of about 24.000 households</u>, selected through a two stage sample design (municipalities/households) from the centralized municipal register (LAC)
- Mixed technique: sequential web-PAPI
 - A self-compiled questionnaire (web) proposed in the inviting letter sent by ISTAT and after, on non respondent households, direct interview with a questionnaire on paper with an interviewer (PAPI)
- In 2017 experimental set up: sequential web/PAPI (MM) with a control single mode (SM) sample PAPI
- □ The selected **sample** of individuals was **linked to an administrative data base** (Archimede Project) through the individual code available from the selection frame to obtain external auxiliary variables



Response rates for ADLs in the SM and MM surveys by geographical area

	Response rates			
Geographical area	SINGLE MODE/PAPI	MIXED MO	DE	
		web	final	
North West	65.9%	32.5%	71.2%	
North East	70.2%	36.0%	73.6%	
Center	68.6%	27.8% 70.2		
South	79.3%	17.7%	79.4%	
Islands	71.3%	17.3%	74.2%	
ITALY	71.0% 26.8% 74.0			

Summary scheme of the experimental context and analyses

Parallel independent samples (SM/MM)	Mixed-mode: Sequential web-PAPI; Control sample: Single mode (PAPI)
Main goal of the analyses	 Evaluation of the impact of the switching from single to mixed mode Evaluation of total non-sampling error components (measurement)
Theoretical context	Instrumental/Counterfactual approaches
Available auxiliary information	Register demo-social covariates
Phases of the analyses (target variables)	 Comparison between the SM and MM samples <u>tests</u> on the <u>differences</u> in the estimates SM and MM <u>study of the total nonresponse bias</u> Analyses on the univariate distributions and multivariate structure of data Assessment of the mode effect, disentangling selection and mesurement (propensity score and instrumental variable)
Phases of the adjustment	 Adjusting for selection effect in the MM design through weighting (standard calibration, fixed mode proportions and propensity score)

The auxiliary variables available for the following analyses and models

Auxiliary mode-insensitive variables in ADL survey at household level:

- <u>Household type</u>: one-component under 55, one-component over 54, couple with children at least one under 25, couple with children without under 25, couple without children, one parent at least one under 25, one parent without under 25, other types
- Higher education level: below/equal/above high school diploma
- <u>Occupation type</u>: Prevalence of: employed, self employed, not in labor age, mixed types
- <u>Municipal type</u>: Metropolitan cities, metropolitan area, other municipalities <2000, 2000-10000, 10000-50000, >50000
- <u>Geographical area</u> (North, Center, South and Islands)
- ✓ Income class: 5 quintiles (€ 11.955, 20.892, 30.028, 46.119)
- <u>Citizenship (nationality)</u>: Italian/Foreign household



The assessment of the introduction of the mixed mode (2)

Analysis of total nonresponse bias – *R*-indicators

 $\square \underline{R-indicators} (Schouten et al., 2011) are based on a measure of the <u>variability of the response propensity</u> and describe how the sample of respondents to a survey reflects the population of interest with respect to certain characteristics
<math display="block">P(a_1) = 1 - 2S(a_2)$

$$R(\rho_X) = 1 - 2S(\rho_X)$$

At national level MM sample deviates less from the representative response with respect to the SM sample – <u>MM sample is more representative</u>

R-indicators in SM and MM samples

		<i>R</i> _Indicator	SM sample	MM sample
response models defin at national lev		Italy	0.812	0.852
rosponso models	response models defined for each	North	0.847	0.840
defined for each		Center	0.752	0.842
geographical area	L	South and Islands	0.840	0.907



The impact of MM on the univariate and multivariate structure of data

Users' intevest is generally the relations among variables, studied through statistical models

What is the impact of data collection design on distributions and/or associative structure of the variables? (*Martin and Lynn, 2011*)

Univariate analysis - impact of mixed-mode design (SM/MM) on the distributions of ADL variables

- Regression models, with the survey variable as the dependent variable and a dummy variable "survey design" as the independent variable
 - ✓ appropriate statistical models and tests to evaluate if the distributions are significantly different

Multivariate analysis - impact of mixed-mode design (SM/MM) on the estimation of models

- Regression models, with interaction effects between "survey design" and auxiliary socio-demographic variables to estimate the association
 - ✓ appropriate statistical models and tests to evaluate the statistically significance of the interaction effects
 - Significant interaction effects would show different relations among structural and target variable depending on the survey design



The impact of MM on the univariate and multivariate structure of data (2)

Results – regression model with independent variable "survey design"

		coeffic	ient	p-va	lue	ANOVA
VARIABLE	Category	Intercept	Survey design	Intercept S	urvey design	p-value
	Sometimes a week	0,500	-0,050	0,000	0,111	
	Once a week	0,204	-0,039	0,000	0,241	
Frequency of seeing friends	Sometimes a month	0,175	-0,139	0,000	0,000	
(Everyday)	Sometimes a year	-0,407	-0,317	0,000	0,000	
	Never	-1,114	-0,165	0,000	0,002	
	No friends	-2,167	-0,281	0,000	0,001	
	NR	-2,411	-0,458	0,000	0,000	0,000
Performing physical activity	Sometimes a week	-0,828	0,189	0,000	0,000	
(NO)	Sometimes a month	-1,643	-0,124	0,000	0,006	
	Sometimes a year	-1,527	-0,248	0,000	0,000	
	NR	-2,588	-0,025	0,000	0,702	0,000
Playing sports, with continuity	Yes	-1,117	0,097	0,000	0,000	
(NO)	NR	-3,835	-0,134	0,000	0,117	0,000
Playing sports, occasionally (NO)	Yes	-1,926	0,013	0,000	0,719	
	NR	-3,312	-0,097	0,000	0,168	0,345
Hospitalized, in last 3 months	Yes	-3,427	-0,009	0,000	0,871	
(NO)	NR	-3,923	-0,184	0,000	0,020	0,061

The impact of MM on the univariate and multivariate structure of data (3)

Results – regression model with interaction effects between "survey design" and auxiliary variables

VARIABLE	Category	Single effect
	Sometimes a week	Sex, Age class, Educational level, Income class, Occupation type, Geographical Area, Municipal type
Performing physical activity	Sometimes a month	Age class, Educational level, Income class, Occupation type, Geographical Area, Municipal type
(NO)	Sometimes a year	Survey design, Age class, Citizenship, Educational level, Income class, Occupation type, Geographical Area, Municipal type
	NR	Age class, Citizenship, Educational level, Municipal type
	Sometimes a week	Sex, Age class, Educational level, Occupation type, Geographical Area, Municipal type
	Once a week	Survey design, Sex, Age class, Educational level, Income class, Occupation type, Geographical Area, Municipal type
Frequency of seeing	Sometimes a month	Survey design, Sex, Age class, Citizenship, Educational level, Income class, Occupation type, Geographical Area, Municipal type
friends (Everyday)	Sometimes a year	Sex, Age class, Citizenship, Educational level, Income class, Occupation type, Geographical Area, Municipal type
	Never	Sex, Age class, Citizenship, Income class, Geographical Area, Municipal type
	No friends	Sex, Age class, Citizenship, Income class, Occupation type, Geographical Area, Municipal type
	NR	Sex, Age class, Educational level, Occupation type, Geographical Area, Municipal type
Playing sports, with continuity	Yes	Survey design, Sex, Age class, Citizenship, Educational level, Income class, Occupation type, Geographical Area, Municipal type
(NO)	NR	Survey design, Sex, Age class, Citizenship, Income class, Occupation type, Municipal type

Results – regression model with interaction effects between "survey design" and auxiliary variables

VARIABLE	Category	Interaction effects: survey design
Performing physical activity	Sometimes a week	Age class, Geographical Area
	Sometimes a month	Sex
(NO)	Sometimes a year	Citizenship
	NR	Geographical Area
	Sometimes a week	Sex, Educational level, Geographical area
	Once a week	Age class, Municipal type
Frequency of seeing	Sometimes a month	Age class
friends (Everyday)	Sometimes a year	Age class, Geographical area, Municipal type
(Lvelyday)	Never	Sex, Age class
	No friends	-
	NR	-
Playing sports, with continuity (NO)	Yes	Age class, Educational level, Geographical area
	NR	Age class, Educational level, Income class, Occupation type, Geographical area, Municipal type

Selection and measurement effects estimated through different approaches

Estimates of selection and measurement effects - Instrumental variable (SM/MM samples)

Variable	Category	Selection effect	Measurement effect		
Reading books	No	0,1478	-0,0727		
in the last	Yes	-0,1767	0,0416		
12 months	NR	0,0288	0,0311		

Estimates of selection and measurement effects - Propensity Score Subclassification (MM sample)

Variable	Category	Weighted Web mean	Web mean	PAPI mean	Selection effect	Measurement effect
Reading books	No	0.485	0.451	0.618	0.034	-0.132
in the last 12 months	Yes	0.432	0.508	0.347	-0.075	0.085
	NR	0.043	0.041	0.035	0.002	0.007

Methods for adjusting selection effect - Weighting methods

- Propensity score, calibration of weights modified through the correction factors (*Rosenbaum and Rubin, 1983 - Vandenplas et al., 2016*)
- Standard calibration on demographic totals
- Calibration on fixed levels of mode proportions (method proposed by *Buelens and Van den Brakel, 2015*), to stabilize the selection effect in repeated surveys, assuming the invariance of measurement effect, with the aim to obtain reliable changes over time
- Assuming the hypothesis of ignorability of the selection effect and absence/stability of measurement effect



The adjustment of selection effect with different methods (2)

Comparison of the estimates deriving from the application of different methods

Methods based on **calibration** on distributions of the same socio-demographic totals (age class, sex, educational level) at geographical area level, but different for other aspects:

- 1) only socio-demographics;
- 2) socio-demographics and <u>observed fixed levels of mode proportions</u> by six municipal typologies;
- 3) socio-demographics and <u>hypothesized fixed levels of mode proportions</u> by six municipal typologies;
- 4) socio-demographics with sampling weights corrected for the web selection effect through correction factors w_k (propensity score)

Estimates of "reading books in the last 12 months" with different methods

A	Catagoria	Estimate (%)				
Variable	Category	Meth. 1	Meth. 2	Meth. 3	Meth. 4	
Reading books	No	59,82	58,88	58,54	59,81	
in the last	Yes	36,51	37,47	37,76	36,46	
12 months	NR	3,67	3,65	3,70	3,73	

Final considerations

- □ For the Aspect of Daily Life survey
 - the introduction of mixed mode has an important impact both on the composition of the sample (and its representativeness) and on several indicators, whose quality seems to be affected by measurement effect which cannot be always easily assessed
 - MM seems to have an impact on simple and complex analyses as well
 - the application of all the presented methods is subject to the validity of the hypotheses underlying all these methods and that need to be verified by the researcher as far as possible
- The set of the analyses presented and applied in a specific survey context can be considered as a **possible checklist**, a **sequence of steps** usable by researchers of other NSIs to carry out an assessment of mode effect in similar situations
- Generally the underlying effort is hardly compatible with the usual resources and the timing of a statistical process: in general situations an accurate planning of the data collection phase is more advisable, in order to limit as far as possible ex-ante the measurement effect, which is the main drawback of the mixed mode



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Thank you for your attention !

