

Center for Economic and Social Research Program on Global Aging, Health, and Policy

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Data Harmonization

December 3, 2019

Gateway to Global Aging Data Team

Outline

- Introduction to harmonization
 - Gateway to Global Aging Data
 - Types of data
 - Basic principles
- Building a metadata library
 - Extracting and indexing metadata
 - Building search engine
 - Communicating data comparability
- Harmonizing phenotype data
 - Pre-statistical harmonization
 - Missing data analysis and imputation
 - Statistical harmonization

Introduction to Harmonization

Gateway to Global Aging Data

The platform for population survey data on aging around the world

The Gateway is a free public resource designed to facilitate cross-national and longitudinal studies on aging using the family of Health and Retirement Studies around the world.



www.g2aging.org

Health and Retirement Studies Around the World

- Republic of Ireland Estonia Israel • • Hungary 🦉 • Italy Poland • Netherlands • Portugal Japan Indonesia Slovenia • Spain Sweden China Switzerland Ghana Croatia Costa Rica • India • Brazil South Korea Russia
 - South Africa

Czech Republic

• Luxembourg

- Scotland
- Northern Ireland

- Bulgaria •
- Cyprus
- Finland
- Latvia
- Lithuania
- Malta
- Romania
- Slovakia
- India

United States

• Austria

Mexico

- Belgium
- Denmark
- France
- Germany
- Greece

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A platform for population survey data on aging around the world

HELP

FAQ

Register

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Q

What the Gateway Offers

- Overview of HRS family surveys
 - Survey questionnaires
 - Flow-charts illustrating questionnaire skip patterns
 - Search engine by keyword or by topic
- Cross-study concordance tables of specific survey topics
- In-depth documentation of cross-study comparability
- Harmonized data
- Interactive graphs and tables
 - Survey statistics
 - Contextual data
- Search of publications based on HRS family surveys

Types of Data

- Metadata
 - Survey questionnaire, data collection protocol
 - Para data time-stamps, interviewer notes
- Microdata
 - Phenotype data survey data, biological markers
 - Genomic data polygenic risk scores
- Contextual data
 - Macro-level data unemployment rates, no of hospital beds
 - Environmental data pollution, temperature
 - Institutional data pension, long-term care policies

Basic Principles

- Accuracy
- Transparency
- Ease of use

With the ultimate goal of advancing science by supporting data users

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- University of Duesseldorf, Germany
- University of Malaysia, Malaysia
- University of Michigan
- University of Stirling, UK
- University of Texas, Medical Branch
- University of Tokyo, Japan
- World Health Organization

Basic Principles – Accuracy

 We work closely with study teams to devise a harmonization plan

 All documentation and programming are reviewed through a standardized quality control process

Survey Measure	Study 1	Study 2	Harmonization Rating
Self-rated health	5 point scale, excellent to poor	5 point scale, excellent to poor	Comparable
Smoking quantity	Packs smoked per day	Packs smoked per week	Can be adjusted
Whether any vigorous physical activity	3 or more time per week	More than 10 minutes per week	Cannot be adjusted
Word recall test	10 word recall	8 word recall	Requires statistical calibration

Basic Principles – Transparency

Concordance Tables

Measure	Study 1	Study 2	Study 3	Study 4	
Word recall	Waves 1 -3: 20 words After wave 4: 10 words	All waves: 8 words	All waves: 10 words	All waves: 10 words	

User Guides



Codebooks





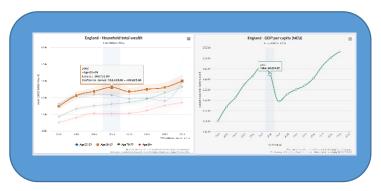
Basic Principles – Ease of Use

Intuitive Search

- Study and time specific
- Across studies
- Across repeated
 observations
- Time point comparisons



Graphs & Tables



Researcher Trainings



Links and Instructions



Building a Metadata Library

Collecting Survey Metadata

Survey metadata include:

- Data collection protocol
- Order of modules and order of questions within modules
- The location of all survey items inside the interview
- How the measure was collected and from whom
- Question text and interviewer prompts
- Answer types and choices and how the values are formatted
- Interviewer notes, time-stamps, and other para-data



Collecting Survey Metadata

• Survey metadata can be obtained using:

Optical character recognition (OCR) from paper versions of survey instruments

SECTIONA DEMOCRAPHICS

	GENERAL DATA		A.7	(Before you were a toilet?	ge ten).	did yo	ur hou	se have	
D.	what day, month, and year were you born? IAY			YES NO RF DK		2			\Box
	EARlll		A.8	(Before you were a rious health probl mal activities for a	em tha month	t affec or mo	ted yo		
A.2 In w	what State/Country were you born?			YES NO RF DK		2			
·	K		A.9	Before you were a any of the followi					
plet LEVE Pr Se Te Pr Ba Ca G	$\begin{array}{c} \text{one} & & & 0 \\ \text{imary} & & 1 \\ \text{econdary} & & 2 \\ \text{econdary} & & 3 \\ \text{reparatory or High School} & & 3 \\ \text{reparatory or High School} & & 5 \\ \text{olege} & & 6 \\ \text{olege} & & 6 \\ \text{olege} & & 6 \\ \text{reparatory or aduate} & & 7 \\ \end{array} \rightarrow \begin{array}{c} \text{Go to A.6} \\ \text{solution} & \text{solution} \\ \text{reparatory or aduate} & & 7 \\ \text{result} & \text{solution} \\ \text{result} & & 9 \\ \text{solution} & \text{solution} \\ \text{result} & \text{solution} \\ \text{result} & \text{solution} \\ \text{result} & \text{result} & \text{result} \\ \text{result} & \text{result} \\ \text{result} & \text{result} \\ $	Ц	Ri Po Ty A	berculosis ieumatic Fever lio phoid Fever serious blow to the ad that made you nt?	YES 1 1 1 1 1 1 1	NO 2 2 2 2 2 2 2	RF 8 8 8 8 8	DK 9 9 9 9 9 9	
GRA	ADE			M	IARITAL	STAT	US		
sag YE NK	you know how to read and write a mes- e? ES	Ш	A.1	0 Currently are you single? married? in a consensual un		2	Go		

Parsing the code of a computer-assisted personal interviewing (CAPI) survey instrument

{ This code was generated by a tool on Wednesday, December 04, 2013 at 3:10:19 AM. } { Colectica® 4.1.3341 Release } { }	<i>.</i>
Changes to this file may cause incorrect behavior and will be lost if the code is regenerated. }	
DATAMODEL _2010Instrument "2010 United States Census Questionnaire"	
INCLUDE "BBlock1.inc.bla"	
TYPE	
TadditionalPeopleCodes =	
(
C1 (1) "Children, such as newborn babies or foster children",	
C2 (2) "Relatives, such as adult children, cousins, or in-laws", C3 (3) "Nonrelatives, such as roommates or live-in baby sitters",	
C4 (4) "People staying here temporarily",	
C5 (5) "No additional people"	
ThouseOwnershipCodes =	
(
C1 (1) "Owned by you or someone in this household with a mortgage or loan? Include home equity	
C2 (2) "Owned by you or someone in this household free and clear (without a mortgage or loan)?",	
C3 (3) "Rented?", C4 (4) "Occurried without payment of cost?"	
C4 (4) "Occupied without payment of rent?"	
5	
FIELDS	
{ Name: Q1 }	
Q1 "How many people were living or staying in this house, apartment, or mobile home on April 1, 2010?" :	Integer
{ Name: 02 }	
Q2 "Were there any additional people staying here April 1, 2010 that you did not include in Question 1?"	: Tadditional
{ Name: Q3 }	
Q3 "Is this house, apartment, or mobile home" : ThouseOwnershipCodes	
{ Name: 04 }	
04 "What is your telephone number? We may call if we don't understand an answer." : Strino[10]	•
	<u>1</u> (

Indexing Survey Metadata

- Collected survey metadata information are indexed into an objectoriented database where all internal components are connected to each other
- This database defines the structure of all survey items

Survey measure 1648
Part of module 298
In survey 83 (conducted in 2012)
★ From of study 3

 The database also allows for instant querying by any indexed dimension of the survey metadata

SELECT all survey metadata TYPE: Question CONTAINING TEXT: "smoking" IN THE TIME PERIOD: 2010-2015

Indexing Survey Metadata

Object-oriented database

🔄 Browse 🥖 Structure 🚊 SQL 🔍 Search 🐉 Insert 🛶 Export 🛶 Import 🗉 Privileges 🥜 Operations 🐹 Triggers

1 \$ > >> Number of rows: 25 \$ Filter rows: Search this table

- Options surveytitle	moduletitle	itemlabel	description a 1	answer_type	answer
SHARE 2017	HC. Health Care	HC097	How much did you pay overall for your nursing home stays in the last twelve months?	String	
SHARE 2017	SP. Social Support	SP001	The next questions are about the help that you may have given to people you k now or that you may have received from people you know.	Enumerated	1 Continue
SHARE 2017	HC. Health Care	HC114	Was there a time in the past 12 months when you needed to see a doctor but could not because of cost?	Enumerated	1 Yes 5 No
SHARE 2017	SP. Social Support	SP007	(Please look at card 27) is there any other family member from outside the household, friend or neighbour who has given you personal care or practical household help?	Enumerated	1 Yes 5 No
SHARE 2017	SP. Social Support	SP013	(Please look at card 27) is there any other family member from outside the household, friend, orneighbour to whom you have given personal care or practical household help? HAVE YOU GIVEN HELP TO OTHERS	Enumerated	1 Yes 5 No
SHARE 2017	EX. Expectations	EX009	(Please look at card 39.) What are the chances that you will live to be age [{Current age rounded up to 5 fold}] or more?	Range	0100
SHARE 2017	EX. Expectations	EX025	(Please look at card 39.) Think ing about your work generally and not just your present job, what are the chances that you will be work ing full-time after you reach age 63?	Range	0100
HARE 2017	EX. Expectations	EX008	(Please look at card 39.)What are the chances that before you retire the government will raise your retirement age?	Range	0100
SHARE 2017	EX. Expectations	EX007	(Please look at card 39.)What are the chances that before you retire the government will reduce the pension which you are entitled to?	Range	0100
SHARE 2017	BR. Behavioural Risks	BR029	(Please look at card {SHOWCARD_ID}) In a regular week , how often do you consume a serving of fruits or vegetables?	Enumerated	1 Every day 2 3-6 times a week 3 Twice a week 4 Once a week 5 Less than once a week
SHARE 2017	BR. Behavioural Risks	BR028	(Please look at card {SHOWCARD_ID}) In a regular week , how often do you eat meat, fish or poultry?	Enumerated	1 Every day 2 3-6 times a week 3 Twice a week 4 Once a week 5 Less than once a week
SHARE 2017	BR. Behavioural Risks	BR027	(Please look at card {SHOWCARD_ID}) In a regular week , how often do you have a serving of legumes, beans or eggs?	Enumerated	1 Every day 2 3-6 times a week 3 Twice a week 4 Once a week 5 Less than once a week
SHARE 2017	FT. Financial Transfers	FT014	(Still think ing about the last twelve months). Is there anyone else inside or outside this household who has given you [or/or/or/or] [your/your/your/ [husband/wife/partner/partner] any financial or material gift or support amounting to [FLDefault[32]] [FLDefault[39]] or more?	Enumerated	1 Yes 5 No
SHARE 2017	CF. Cognitive Function Module	CF113	A little while ago, I read you a list of words and you repeated the ones you could remember. Please tell me any of the words that you can remember now?	Enumerated	1. ^FLMovies[17] 2. ^FLMovies[18] 3. ^FLMovies[20] 5. ^FLMovies[20] 5. ^FLMovies[21] 6. ^FLMovies[22] 7. ^FLMovies[24] 8. ^FLMovies[24] 10. ^FLMovies[25] 10. ^FLDofault[67]
SHARE 2017	CF. Cognitive Function Module	CF114	A little while ago, I read you a list of words and you repeated the ones you could remember. Please tell me any of the words that you can remember now?	Enumerated	1. ^FLMovies[27] 2. ^FLMovies[28] 3. ^FLMovies[29] 4. ^FLMovies[30] 5. ^FLMovies[31] 6. ^FLMovies[32] 7. ^FLMovies[33]

Constructing a Browse/Search Tool

Researchers should be able to easily browse through the survey metadata while understanding the study, timing, and type of the survey

Study O	verview	Core Interview		End of Life Intervie	w	Life History	Healt	h Assessment	Self-Co	mpletion
	HRS	MHAS	ELSA	SHARE	CRELES	KLoSA	JSTAR	TILDA	CHARLS	LASI
	United States	Mexico	England	20+ European Countries and Israel	Costa Rica	Korea	Japan	Ireland	China	India
1992-93	HRS W1									
1772 70	AHEAD 1993 W1									
1994-95	HRS W2									
1774-75	AHEAD 1995 W2									
1996-97	HRS W3									
1998-99	HRS W4									
2000-01	HRS W5	MHAS W1								
2002-03	HRS W6	MHAS W2	ELSA W1							
2004-05	HRS W7		ELSA W2	SHARE W1	CRELES W1					
2006-07	HRS W8		ELSA W3	SHARE W2	CRELES W2	KLoSA W1	JSTAR W1			
2008-09	HRS W9		ELSA W4		CRELES W3	KLoSA W2	JSTAR W2			
2010-11	HRS W10		ELSA W5	SHARE W4	CRELES W4	KLoSA W3	JSTAR W3	TILDA W1	CHARLS W1	
2012-13	HRS W11	MHAS W3	ELSA W6	SHARE W5	CRELES W5	KLoSA W4	JSTAR W4	TILDA W2	CHARLS W2	
2014-15	HRS W12	MHAS W4	ELSA W7			KLoSA W5		TILDA W3	CHARLS W4	
2014-15	UAS HRS W1	IVIMAS VV4	ELSA W/	SHARE W6		KLOSA WS		TILDA W3	CHARLS W4	
2016-17	HRS W13		ELSA W8	SHARE W7		KLoSA W6		TILDA W4		LASIW

Constructing a Browse/Search Tool

Researchers should also be able to intuitively and quickly search through survey metadata to find survey items which are important for their research

Search all surveys b		Keyword	Source		Years
keyword	subtopic	smoking	18 of 28 selected	2010	2016
			Search		
A / Search results 0					
Showing 1 - 50 of 409 result(s):					50 • results per page
Filter by source		AS ELSA SHARE CRELES KLoSA JST	AR TILDA CHARLS LASI SAGE HAALSI HAGIS NICOLA	ELSI HART MARS	•
Filter by year	Ali 2010/1	2012/3 2014/5 2016/7			Þ
LABEL	SURVEY	MODULE		DETAILS	
			Description: Parents/Guardians Smoke		
			Text: Did your parents or guardians smoke during your childhood?	?	
MB104	HRS 2010	B. Demographics	Response type: Enumerated		
			Responses: 1 Yes, one of them		
			2 Yes, both 5 No, none of them		

Constructing a Browse/Search Tool

It is also possible to identify survey items specific to common research topics and domains from which users can see all relevant survey items

Topics



Cognition
 Testing conditions

Healthcare Utilization & Insurance
 Hospital stay

Harmonizing Phenotype Data

Data Harmonization

- 1. Pre-statistical harmonization
- 2. Missing data and imputation
- 3. Statistical harmonization based on Item Response Theory

Pre-statistical Harmonization

Harmonized variables are research-ready variables which easily allow researchers to conduct analysis by pooling data from multiple waves of a survey or from multiple surveys

- Variables are defined identically across all waves and surveys
- Each dataset combines all available waves from each study; each individual is one record

	St	udy 1			Stu	dy 2	
Person ID	R1AGE	R2AGE	R3AGE	Person ID	R1AGE	R2AGE	R3AGE
10000	52	54	56	1000	71	73	75
10001	59	61	63	1001	60	62	64

Pre-statistical Harmonization

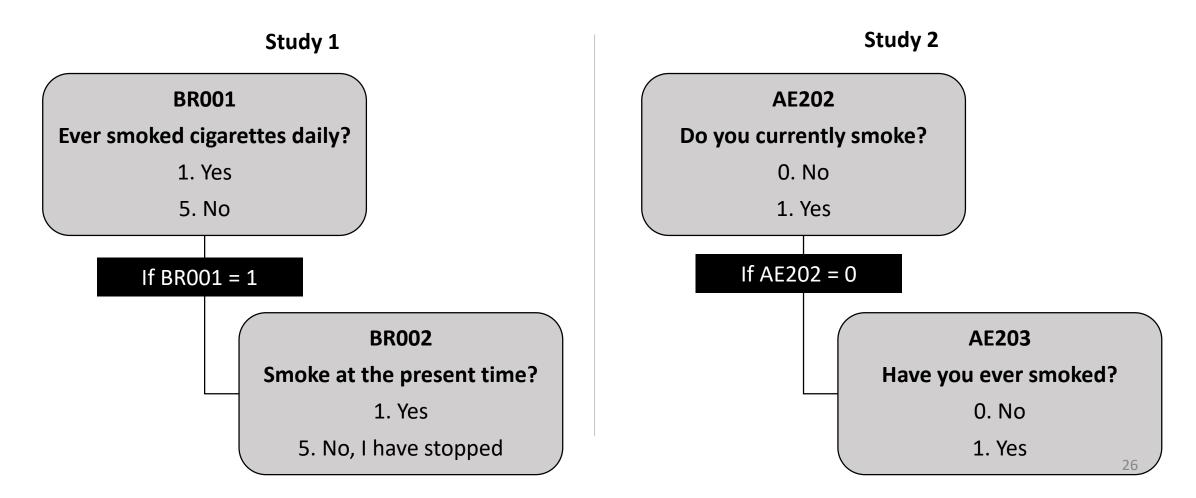
Harmonized variable names ensure ease of use and transparency

- All variables use intuitive variable names, e.g. r1work whether the respondent is currently working in wave 1
- Study specific variable names are used to indicate significant interstudy differences: e.g. RwVGACT_C – whether the respondent does any vigorous physical with different question wording

	St	udy 1			Stud	ly 2	
Person ID	R1VGACT	R2VGACT	R3VGACT	Person ID	R1VGACT_C	R2VGACT_C	R3VGACT_C
10000	1. Yes	0. No	1. Yes	1000	0. No	0. No	0.No
10001	1. Yes	1. Yes	1. Yes	1001	1. Yes	1. Yes	0.No

Building Harmonized Variables

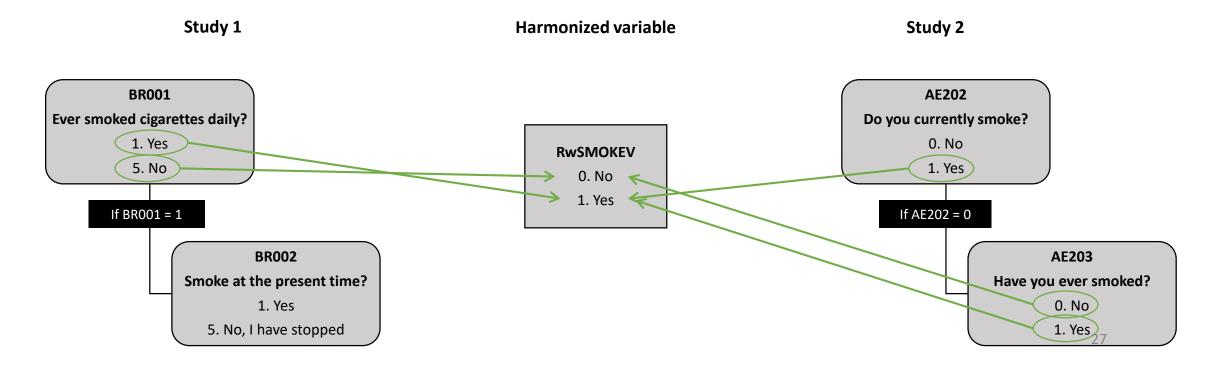
Harmonized variables have been built to account for any survey skip pattern



Building Harmonized Variables

Harmonized variable name: RwSMOKEV Harmonized variable label: Whether the respondent ever smoked Harmonized variable codes:

- 0. No, the respondent has never smoked
- 1. Yes, the respondent has smoked



Documenting Harmonized Variables

Each harmonized dataset is accompanied by its own codebook.

USC Dornsife Genter for Economic and Social Research Program on Global Aging, Health, and Policy	
	Harmonized SHARE Documentation
	VERSION E (2004-2018), October 2019
	Sandy Chien, Drystan Phillips, Marieta Valev, Jenny Wilkens, and Jinkook Lee
	We greatly appreciate support from the National Institute on Aging (Ror AGogolics, RC: AGogélicia, Ros AGo4gocs)
	gaaging.org

- Introduces the harmonization project and study
- Overviews survey timing, survey design, and sampling framework
- Discusses weighting and imputation
- Details specifics of harmonization process
- Divides variables into sections based on research domain

SHARE VERSION AND ACKNOWLEDGMENT	
SHARE VERSION AND ACKNOWLEDGMENT	2
WHAT'S NEW IN VERSION E OF THE HARMONIZ	ED SHARE? 3
1. INTRODUCTION AND OVERVIEW 7	
1.1. Gateway to Global Aging Data	
1.2. Units of Observation	
1.3. Data File Structure	
1.4. Variable Naming Convention 1.5. Missing Values, Nonresponse, and Imputation	
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2. WEALTH AND INCOME VARIABLES 13	
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Currencies, timing, and exchange rates	
2.3. Differences between Harmonized SHARE and	RAND HRS
3. STRUCTURE OF CODEBOOK 18	
4. DISTRIBUTION AND TECHNICAL NOTES	21
5. DATA CODEBOOK 22	
Section A: Demographics, Identifiers, and Weights	
Section B: Health	
Section C: Health Care Utilization and Insurance	
Section D: Cognition Section E: Financial and Housing Wealth	340
Section F: Income and Consumption	
Section G: Family Structure	
Section H: Employment History	
Section I: Retirement and Expectations	
Section J: Pension	
Section K: Physical Measures	
Section M: Stress	
Section O: End of Life Planning	
6. REFERENCES 711	

Documenting Harmonized Variables

Health Behaviors: Smoking (Cigarettes) Summary statistics ٠ Wave Variable Label Type R1SMOKEV rlsmokev:wl r smoke ever Categ R2SMOKEV r2smokev:w2 r smoke ever Categ for each set of variables Categ Categ R4SMOKE1 r4smokev:w4 r smoke ever 5 R5SMOKEV r5smokev:w5 r smoke ever Descriptive Statistics Variable Mean Std Dev Minimum Maximum 0.00 0.00 0.00 0.00 1.00 1.00 1.00 1.00 R1SMOKEV 0.47 0.50 36674 57083 65429 0.47 0.50 R2SMOKEV R4SMOKEV 0.50 R5SMOKE) Categorical Variable Codes Tabulations of all coded values ٠ R1SMOKEV R2SMOKEV R4SMOKEV R5SMOKEV .d:DK .m:Missing 148 501 1092 809 .r:Refuse 16013 19289 0.no 30395 34390 14278 17385 26688 31039 1.yes Details about variable creation and any How Constructed . RwSMOKEV indicates whether the respondent reports ever having smoked cigarettes, pipes, or cigars daily for a period of at least one year. The answer to the respondent's first ever-smoked daily question is carried forward in subsequent waves. A code of 0 indicates that the respondent has never smoked daily. A assumptions made in the creation code of 1 indicates that the respondent has ever smoked daily. When respondents don't know, refuse to answer, or are missing, RwSMOKEV is assigned special missing values .d, .r, .m, respectively. RwSMOKEV is set to plain missing (.) for respondents who did not respond to the current wave. Cross Wave Differences in SHARE Highlights of any intra-study differences o differences known ٠ Differences with the RAND HRS In the SHARE, respondents are asked whether they have ever smoked daily for a period of at least one year. In the HRS, respondents are asked whether they have ever smoked (regardless of whether the smoking was daily and not given a definitive period). Consequentially, RwSMOKEV in the Harmonized SHARE captures a different concept them RwSMOKEV in the RAND HRS. This difference also affects RwSMOKEN in the Highlights of any inter-study differences . Harmonized SHARE because of the question routing explained above. Only SHARE respondents who answered that they have ever smoked daily for a period of at least one year were asked whether they smoke currently. In the HRS, all respondents who reported that they had ever smoked (regardless of whether the smoking was daily for a specific period) were directed to the guestion ever smoke currently. These two sets of measures should not be considered exactly comparable to the correlating RAND HRS measures. In the HRS, the question about whether a person ever smoked daily is only asked at the respondent's first interview. For each respondent the answer to such question is carried forward in subsequent waves List of all the variables from the • SHARE Variables Used Wave 1: ever smoked daily BR002 smoke at the present time originating dataset used in the creation Wave 2: BR001 ever smoked daily BR002 smoke at the present time Wave 4: of the harmonized variable BR001 ever smoked daily BR002 smoke at the present time Wave 5: BR001 ever smoked daily BR002 smoke at the present time

- Complete case analysis is the default for most statistical software
- Excluding observations with missing values is potentially a huge loss of information (large standard errors; imprecise estimators)
- Excluding observations with missing values also potentially introduces large biases (people with lower levels of cognition are more likely to have missing values in cognitive tests)

Person ID	Gender	Age	Income Level	Education	Marital Status	Word Recall Score	INCLUSION
1	Male	70	High	HS graduate	Married	8	YES
2	Female	55	High	College grad	Married	10	YES
3	Male	89	Low	HS graduate	Widowed	•	NO

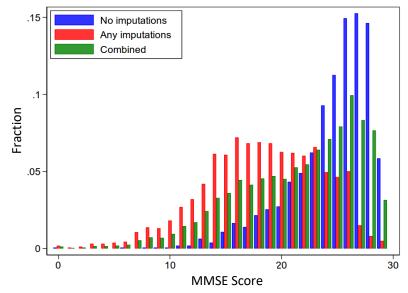
- Imputed values replaces missing values with draws from their (conditional) distribution, and thereby creates a complete data set for researchers
- Imputation is often economically efficient and scientifically preferable because it is done only once (individual researchers do not have to do it, and all researchers work with the same data)

Person ID	Gender	Age	Income Level	Education	Marital Status	Word Recall Score	Word Recall Score w/ Imputations	NCLUSION
1	Male	70	High	HS graduate	Married	8	8	YES
2	Female	55	High	College grad	Married	10	10	YES
3	Male	89	Low	HS graduate	Widowed		3	YES

Harmonized imputation method

- Estimates the joint distribution of the variables in each dataset in the presence of missing data
- Assigns "don't know" values for many cognitive tests as 0 values
- Estimates a regression model which specifies the conditional distribution of the variable to be imputed as a function of the regressors.
- Imputes missing values using pseudo-random draws from the conditional distribution

Person	Gender	Age	Value	Value w/ Imputations
1	Male	60	20	20
2	Male	60		20
3	Male	72	10	10

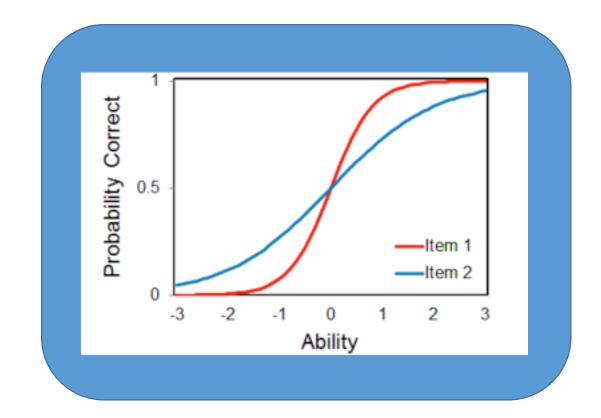


Researchers are provided with indicators of whether values were imputed and what level of information was used in the imputation procedure

Value	R1FMO
0. Not imputed	2833
1. DK – assigned 0 value	273
2. Not assessed - imputed	91
3. Refused - imputed	25
4. Missing - imputed	2

Item Response Theory

- A general approach to data analysis relating responses to underlying traits
- Many related statistical models
- Broadly contained within general latent variable framework
- Developed in fields of Educational and Psychological Assessment (1930-1970's)
- Continually refined methods
- Broad applications in social and health sciences



Working with psychometricians we develop harmonized domain-specific models

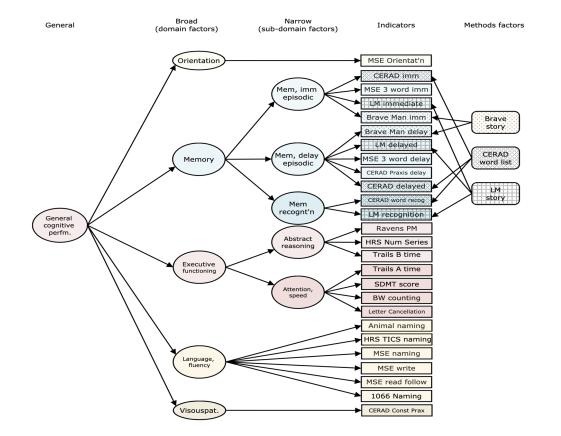
- Classify items according to measurement
 - Domain
 - Modality (performance, self-rated, informant-rated, expert rating)
 - Response type

• Identify reference population and sample(s)

Date Naming Domain: Time and Place Orientation Modality: Performance Response Type: Boolean

Reference population: Clinician diagnosed individuals with moderate and severe cognitive impairment as part of study ABC

• Assess cross-validation of the measurement model using IRT



Model	Descriptor
Single domain models	
Orientation	Good
Memory-Episodic-Immediate	Perfect
Memory-Episodic-Delayed	Good
Memory-Episodic-Recognition	Perfect
EF-Abstract-Reasoning	Perfect
EF-Attention-Speed	Good
Language	Good
Visuospatial	Perfect

• Rank and/or classify individuals

Episodic Memory Impairment Cut Points None: >=16 Mild: 13-15 Serious: 11-12 Severe: <= 10

• Co-calibrate across studies and time

Study	Score Equivalence									
Study 1	1	2	3	4	5	6	7		8	
Study 2	1		2		3	4	5	6	7	8
Study 3	1	2	3			1	5	6	7	8

Thank You

National Institute on Aging, NIH

(R01 AG030153, RC2 AG036619, R03 AG043052, R24 AG048024)