Revisiting the Measurement and Dimensionality of Political Knowledge: Evidence from Seven European Countries

William Allen william.allen@politics.ox.ac.uk

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Kristoffer Ahlstrom-Vij k.ahlstrom-vij@bbk.ac.uk

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Abstract

Knowledge matters for the political preferences people form, and the choices they make. Despite the recent growth of experimental work, however, the bulk of research on the role of knowledge in politics still uses observational data, where researchers typically rely on batteries of general political knowledge questions to distinguish more informed voters from less informed ones. When used to explain outcomes across policy-specific domains, this practice invokes what we call a generalist assumption: possessing general knowledge is diagnostic of holding issue-specific knowledge (and vice versa). As the list of geographical areas and issue domains of interest to political scientists grow, the weight put on this assumption increases significantly; so, is it warranted? Using 2018 survey data from seven European countries (Germany, Hungary, Poland, Romania, Spain, Sweden, UK) that, unusually, includes knowledge questions about both general politics as well as about EU immigration (N = 10,749), we combine several approaches to test the core expectations of the generalist assumption: First, exploratory and confirmatory factor analyses suggest that the two sets of questions are plausibly unidimensional. Second, after constructing Item Response Theory (IRT) knowledge scales from the items, we demonstrate how both scales also display similar associations with key respondent features (gender, age, and education) that align with prior expectations about who typically holds more political knowledge, and with two reference sets containing established measures of general political knowledge, from BES (Wave 17) and ANES (2019 Pilot study). Finally, in investigating whether the evidence uncovered for the generality assumptions goes beyond the specific issue of immigration, we show that the estimated marginal mean level of political as well as immigration knowledge by education, gender, age and (where available) income exhibit similar patterns with independent scales measuring knowledge on public health and climate change. We conclude that the preponderance of evidence points to the generalist assumption. However, we caution against deploying it uncritically and offer our code as a means of enabling others to conduct similar stress-tests.

Keywords: political knowledge; immigration; public health; climate change

1 Introduction

Knowledge matters for politics, and yet citizens are generally un- or misinformed about politics (Fowler and Margolis 2014). This presents a problem not only for the integrity of their political voice but also for representation: If an electorate making choices based on false beliefs is more likely to express political preferences that they might not have held, had they been more informed, then when political representatives cater to the public's stated preferences, the people expressing those preferences are not well-represented (Bartels 1996; Delli Carpini and Keeter 1996). Indeed, as a cognitive division of labour is at the very heart of representative democracy, representatives who uncritically mirror the political voices of the represented run the risk of operating in bad faith.

For these reasons, political scientists are rightly concerned with the relationship between knowledge and political opinion (Kuklinski et al. 2000; Munger et al. 2020). A large body of work—largely experimental in nature—has demonstrated how citizens across diverse contexts respond to factual interventions in predictable ways (Walter et al. 2020), notably by moving their beliefs towards more factually correct positions (Carnahan and Bergan 2021; Guess and Coppock 2018; Porter and Wood 2021). Yet existing evidence on whether and to what extent such knowledge shifts attitudes and preferences in particular directions is less clear, particularly when it comes to specific issues such as immigration which evoke strong partisan cleavages (Blinder and Schaffner 2019; Grigorieff, Roth, and Ubfal 2020; Hopkins, Sides, and Citrin 2018). Moreover, bespoke experimental designs, while highly internally valid, are due to resource constraints often not an option for many researchers and projects, especially large scale comparative ones.

What follows therefore focuses on parallel developments in information effects research which typically uses existing observational data to model to what extent holding different levels of knowledge matter for political behaviours, including vote choice (Bartels 1996;

^{*}Corresponding author; Email: william.allen@politics.ox.ac.uk

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Blais et al. 2009; Delli Carpini and Keeter 1996; Oscarsson 2007) as well as preferences and attitudes on specific issues (Althaus 2003; Ahlstrom-Vij 2021). However, political attitudes encompass a wide range of specific issues: immigration, climate, welfare, foreign policy, to name but a few. In order to make judgments about the role of information across such domains, an assumption is typically made in the information effects literature: people who are knowledgeable in one area of politics will be knowledgeable in others as well (e.g., Zaller 1986 and 1992; Delli Carpini and Keeter 1996). We call this the generalist assumption. The attraction of that assumption should be obvious: if it holds, researchers can use a general knowledge scale to measure the extent to which people are politically informed across specific domains, even in the absence of items directly concerned with the particular topics represented within those domains.

Yet, despite its methodological centrality, this assumption has remained largely untested (although see Delli Carpini and Keeter 1996 for an early exception that we will be returning to below), partly due to the lack of available survey instruments which contain both general and issue-specific factual items. In response, we present analysis using data from a cross-national European survey that, unusually, contains data on respondents' knowledge both about general politics and on a specific issue—in this case, immigration—as well as their attitudes on that issue (Meltzer et al. 2019). By applying several methods to assess the dimensionality and construct validity of these different types of knowledge questions, we conclude that the generalist assumption is defensible but should not be invoked uncritically. Instead, researchers using available knowledge questions on observational surveys would be well-served by stress-testing their conclusions using other data where available. Finally, in line with Open Science principles, we publicly offer our approach (its code and method) as a template for others.¹

2 The Generalist Assumption about Political Knowledge

The generalist assumption can be traced back at least to John Zaller's analysis of information items in the 1985 National Election Study Pilot, where he argued that "political information is a relatively general trait that can be effectively measured with a general-purpose information scale" (Zaller 1986, 2). This conclusion subsequently informed his landmark study on public opinion, where he noted that he was "assuming that persons

¹See https://github.com/ahlstromvij/REMINDER_project. More broadly, we hope our approach enables other researchers to replicate our analyses for their own purposes in so far as they either (a) have data sets with different knowledge items and want to evaluate for dimensionality in order to decide on what scale to use; or (b) are collecting data and want to run a pilot with different items, use our approach, and then potentially save funds and time (and respond to quality concerns relating to survey attentiveness) on only collecting on the items that this analysis suggests that they need (e.g. by skipping domain specific items if not needed).

who are knowledgeable about politics in general are habitually attentive to communications on most particular issues as well" (Zaller 1992, 43). That same assumption is investigated and defended by Delli Carpini and Keeter (1996), and relied upon in both Bartels's (1996) and Althaus's (2003) highly influential work on so-called information effects, i.e. differences between actual preference reports and modeled estimates of the preferences reports respondents likely would have given, had they been fully informed.

In operationalizing the relevant notion of informedness, Bartels relies upon interviewer ratings of respondents' "general level of information about politics and public affairs" (1996, 203) while Althaus uses the type of general scales developed by Delli Carpini and Keeter. Subsequent work across a variety of geographical contexts including the US (Ahlstrom-Vij 2021), Denmark (Bhatti 2010; Hansen 2009), Sweden (Oscarsson 2007) and Canada (Blais et al. 2009) follows the norms set by these early studies by using either a variety of general knowledge items or a scale of both general and specific campaign items alongside interviewer ratings of the respondent's knowledge. As such, the generalist assumption has come to carry an increasingly heavy burden: not only is the catalogue of geographical contexts in which researchers look for information effects expanding, but so also is the list of the specific political and policy issues that feature either directly or indirectly in models of vote choice in a range of political circumstances.

At the same time, there has been a growth in experimental work on the role of knowledge and information on specific policy matters, one such area being immigration. For example, providing corrective information about the levels and impacts of immigration appears to reduce misconceptions and change attitudes (Grigorieff, Roth, and Ubfal 2020), as well as policy preferences in some circumstances (Blinder and Schaffner 2019; Facchini, Margalit, and Nakata 2022), although the presence and direction of these effects is not consistent (Jørgensen and Osmundsen 2020). Clearly, further work is needed, but if only owing to resource constraints, much of that research will need to be observational rather than experimental, in which case researchers will in many cases need to rely on available general knowledge items even when concerned with specific policies issues. This puts further pressure on the generalist assumption.

So, does that assumption hold? That is the question we set out to answer below, initially using immigration knowledge and general political knowledge as our test case, before moving on to a wider set of issue-specific knowledge scales. If the assumption holds, we expect that different methods of analysis (which we describe later) will converge to support its core expectations: that (1) relevant pattern of observed variables can plausibly be represented by a single latent trait (factor) in factor analysis; (2) regressing the knowledge scales on demographic variables will produce associations that one would expect from prior work into the determinants of political knowledge, where those associations will also be in line with patterns found in established measures of political knowledge (in our case, from BES and ANES); and, in terms of generalisability, (3)

the estimated marginal mean level of general political knowledge by education, gender, age and (where available) income should exhibit the same pattern not only on the two scales above, but also on independent scales measuring knowledge on other, issue-specific matters, namely: public health and climate change.

3 Methods and Data

3.1 Dataset

We use a subset of online survey data collected as part of the REMINDER (Role of European Mobility and Its Impacts in Narratives, Debates and EU Reforms) project (Meltzer et al. 2019).² Unlike many existing datasets, the survey component of the REMINDER project is both cross-national and includes general political knowledge questions as well as immigration-specific ones. Therefore, the data set is uniquely suitable for testing the generalist assumption as it is implicitly invoked in information effects research, as per the previous section. Moreover, as the project covers seven European countries, the data also allow us to examine any geographic variation—between Germany, Hungary, Poland, Romania, Spain, Sweden, and the UK—thereby testing if the generalist assumption holds across geographical contexts, or whether previous results on unidimensionality are possibly artifacts of particular national profiles, e.g. the US, where early studies on the unidimensionality of political knowledge took place.

The original REMINDER study used a panel design comprising three waves across 2017-18, whose sampling procedures used quotas by age, gender, and region (at NUTS2 level) to approximate representativeness for each country's adult population. We use data from the second wave (collected between June 6-July 16, 2018) because our outcomes of interest were only asked in that wave alongside the two sets of knowledge questions. Our sample covers 10,749 respondents, with a breakdown of demographic details appearing in Table 1. We conducted multiple imputation for missing values (1.3% of values across all variables used) using the Hmisc package in R (Harrell Jr. et al. 2020).

3.2 General and Specific Knowledge

To measure general political knowledge, we identify correct true/false responses to the following statements:

- (G1) "Switzerland is a member of the EU" (false);
- (G2) "every country in the EU elects the same number of representatives to the European Parliament" (false); and

²Full details about the dataset, including its documentation and questionnaire design, are available at https://doi.org/10.11587/LBSMPQ.

Table 1: Demographic details of the sample

Nationality	Germany	Hungary	Poland	Romania	Spain	Sweden	UK
Mean age	54	46	48	44	48	52	54
Male	914	747	788	703	872	731	800
Female	841	712	717	662	756	689	817
ISCED 0	2	6	2	1	3	3	24
ISCED 1	2	4	6	12	34	37	14
ISCED 2	424	216	22	35	72	164	420
ISCED 3	561	467	535	374	96	476	582
ISCED 4	205	170	168	118	480	180	48
ISCED 5	128	121	4	52	270	239	73
ISCED 6	192	274	177	518	473	205	313
ISCED 7	227	180	562	241	164	95	128
ISCED 8	14	21	29	14	36	21	15
N	1755	1459	1505	1365	1628	1420	1617

(G3) "[NAME OF THE HEAD OF GOVERNMENT] belongs to the [NAME OF CORRECT PARTY]", depending on the survey country.

Then, to measure specific migration knowledge, we identify correct true/false responses to the following statements:

- (I1) "the free movement of persons is a fundamental right guaranteed by the EU to its citizens" (true);
- (I2) "Greece is part of the Schengen Area" (true);
- (I3) "in 2015, Afghans have been the largest group of people that applied for asylum in the EU" (false); and
- (I4) "in 2015, asylum in the EU was more frequently granted to Syrians than any other nationality" (true).

3.3 Methods

In evaluating the generalist assumption, we apply a three-step approach that expands on that taken by Delli Carpini and Keeter in their early work on knowledge scales, and is in our case also line with recent interest in open science by making available all code used for the analyses. First, we use a combination of exploratory and confirmatory factor analysis to investigate the dimensionality of the political knowledge response data. Second, we use regression analysis to investigate construct validity, i.e. whether any latent traits identified exhibit the type of correlations with other, demographic traits that you would expect if they corresponded to political knowledge. To do this, we use Item Response Theory (IRT) to create and validate general and immigration-specific knowledge scales

Table 2: EFA of General (G) and Immigration (I) Knowledge Questions

Item	Factor 1	Factor 2	Factor 3	Factor 4
G1	0.121	0.139	0.21	0.529
G2	0.13	0.207		0.456
G3			1	
I1		0.708		
I2		0.632		
I3	1.002			
I4	0.385	0.378	0.161	-0.237

based on the batteries of questions available in the data set (de Ayala 2009; DeMars 2010). We also use two additional reference datasets from BES (Wave 17) and ANES (2019 Pilot study), in order to calculate estimated marginal mean level of general political knowledge by education, gender, age and (where available) income, and investigate whether these exhibit the same patterns as the two scales above. Third, in order to test whether the evidence thereby uncovered for the generalist assumption, we also compare the patterns on our general and immigration knowledge scales with two, independent scales measuring knowledge on other, issue-specific matters, namely: public health and climate change.

4 Empirical Analysis

4.1 Step 1: Factor analysis

First, we combine all six items from both knowledge batteries, and run a parallel analysis to get an initial estimate of the likely number of factors involved. Such a parallel analysis suggests four factors in our case. We then apply exploratory factor analysis (EFA) set to four factors. As can be seen from Table 2, three of the immigration items (I1, I2, and I4) loaded well onto one factor (Factor 2) while two of the general items (G1 and G2) loaded well onto another (Factor 4). A further parallel analysis of the three general items (G1-3) suggested one factor, while a parallel analysis for the three immigration items does the same, if excluding I3. This offers some initial evidence of two separate factors, corresponding to general knowledge and immigration knowledge, respectively.³

Probing further, in the manner that Delli Carpini and Keeter (1996) also do, we then use confirmatory factor analysis (CFA) to compare the fit of four models: two unidimensional ones and two two-dimensional ones, with the first one in each pair containing all six items and the second one excluding one of these (G3, i.e., the one that failed to load well onto Factor 4 in the above EFA). As shown in Table 3, the model with the best fit

³We do not look at scale reliability as measured by, e.g., Chronbach's alpha, since such reliability can be high even given multidimensionality (Fabrigar and Wegener 2012).

Table 3: CFA fit measures of General (G) and Immigration (I) Knowledge Questions for unidimensional (1D) and two-dimensional (2D) models

Measure	1D, 6 items	1D, 5 items	2D, 6 items	2D, 5 items
RMSEA	0.046	0.046	0.031	0.011
CFI	0.975	0.982	0.990	0.999
TLI	0.958	0.964	0.981	0.998
AGFI	0.970	0.964	0.985	0.996

is two-dimensional and with five items, but the unidimensional models also show very good fit. For context, Root Mean Square Error of Approximation (RMSEA) values of less than 0.05 and 0.01 usually are taken to indicate good and very good fit, respectively (Andrews 2021). A Comparative Fit Index (CFI) or Tucker-Lewis Index (TLI) greater than 0.95 (Dima 2018), and an Adjusted Goodness-of-fit Index (AFGI) greater than 0.9, are typically taken to suggest good fit (Baumgartner and Hombur 1996).

Reading across these two sets of analyses, the EFA results suggests that the knowledge questions are potentially tapping into distinctive dimensions – which is not surprising, given that they were intended to measure different types of knowledge. At the same time, while the CFA results indicate that the two-dimensional models display superior fits, the fact that the unidimensional models also exhibit very good fit suggests that the generalist assumption is not obviously misguided at least at this stage.

4.2 Step 2: Construct validation

Factor analysis enables us to investigate the extent to which patterns in correlations between items can be plausibly explained with reference to a certain number of factors, or underlying traits. However, such analysis does not speak to what those traits are, i.e. to the matter of construct validity. For this reason, we now examine the extent to which the correlations between each of the two knowledge scales and a variety of demographic variables appear the way we would expect if these scales were measuring forms of knowledge—and, moreover, whether the correlations exhibited by the two scales are similar, as they should be, if the generalist assumption holds.

To that end, we first create two knowledge scales using IRT modelling (de Ayala 2009; DeMars 2010). IRT models are used to measure latent traits assumed to fall on a continuous scale. Values on that scale are usually referred to by way of the Greek letter θ (theta), and taken to range from $-\infty$ to $+\infty$, with a mean of 0 and standard deviation of 1. This means that, while the individual θ value ascribed to any particular respondent has no intrinsic meaning, it can nevertheless be interpreted relative to an estimated population mean. Having made sure that the two sets of items satisfy standard assumptions for an IRT model (unidimensionality, local independence, and acceptable model fit), we then

regress these scales on gender, age, and education, across the entire data set (i.e., not segmenting by nationality). The first two variables are self-explanatory. In the data set, the education variable takes on a value from 0 to 8, corresponding to the nine International Standard Classification of Education (ISCED) levels of education, from early childhood education (0) to doctorate (8). Details about the resulting models can be found in Table 4.

Table 4: Regression models (OLS)

	General knowledge	Immigration knowledge
(Intercept)	$-0.842 (0.030)^{***}$	$-0.836 (0.028)^{***}$
Male	$0.230 (0.014)^{***}$	$0.239 (0.013)^{***}$
Age	$0.009 (0.000)^{***}$	$0.008 (0.000)^{***}$
Education	$0.058 (0.004)^{***}$	$0.072 (0.004)^{***}$
\mathbb{R}^2	0.084	0.095
$Adj. R^2$	0.083	0.095
Num. obs.	10749	10749

^{***}p < 0.001; **p < 0.01; *p < 0.05.

As can be seen from Table 4, the coefficient values in both cases cohere with what we know about the relationship between political knowledge and demographics: men tend to have more political knowledge than women (vanHeerde-Hudson 2020; Plutzer 2020), and the educated more political knowledge than the uneducated (Hebbelstrup and Rasmussen 2016). It also seems reasonable to assume that political knowledge increases with age, as indeed is the case on both models. Moreover, as can also be seen from the table, the associations are similar between the two scales, which speaks in favour of unidimensionality, in that it is consistent with both scales measuring the same fundamental underlying trait.

In a further test of construct validation, we also compare the two scales developed here, with two established scales of political knowledge, from the BES and ANES, respectively. In the case of BES, we use Wave 17 of the 2014-2023 British Election Study Internet Panel (Fieldhouse et al. 2020) (N=34,366). That data set contains the following knowledge items, which together form a scale satisfying standard assumptions for an IRT model (as above):

(BES1) No-one may stand for parliament unless they pay a deposit (True)

(BES2) The Liberal Democrats favour a system of proportional representation (True)

(BES3) MPs from different parties are on parliamentary committees (True)

(BES4) The number of members of parliament is about 100 (False)

The ANES data set used is the 2019 Pilot Study (ANES 2019) (N = 3,000), which contains the following knowledge items:

(ANES1) What job or political office is now held by John Roberts? (Correct: Chief Justice of the US Supreme Court)

(ANES2) What job or political office is now held by Angela Merkel? (Correct: Chancellor of Germany)

(ANES3) For how many years is a United States Senator elected – that is, how many years are there in one full term of office for a U.S. Senator? (Correct: 6 years)

(ANES4) On which of the following does the U.S. federal government currently spend the least? Foreign aid; Medicare; National defense; Social Security (Correct: Foreign Aid)

These items, too, form a scale satisfying standard assumptions for an IRT model.

As the education variables differ between the REMINDER data set used above, and the BES and ANES data set (the latter two use standard UK/US education categories rather than ISCED), we consider the best way to compare the associations across these data sets being to plot and compare patterns of marginal means, which we do in Figure 1.

Looking at Figure 1, we can see, first, that the estimated marginal means for each of the three demographic variables are very similar across the general and immigration specific knowledge scales (top two rows). This is consistent with the evidence from previous sections for the generalist assumption – if both scales tap into the same underlying trait, then we should expect the marginal means to exhibit similar patterns across these variables, as indeed they do. Second, the fact that our two scales moreover exhibit similar patterns in their marginal means to the two established scales from BES and ANES (bottom two rows) both offer further evidence for the generalist assumption, and for construct validity, i.e., for the scales tapping into a form of knowledge in particular.

4.3 Step 3: Stress-testing the generality of the generalist assumption

So far, we have seen evidence for the generalist assumption in the context of general political knowledge versus immigration specific knowledge. This evidence does not rule out that there is something unique about immigration specific knowledge in particular, and that the generalist assumption as such would not generalize to other, issue-specific areas. For this reason, in our third and final step, we compare the marginal means of our two scales to that of two, separate issue-specific scales: one relating to public health, and

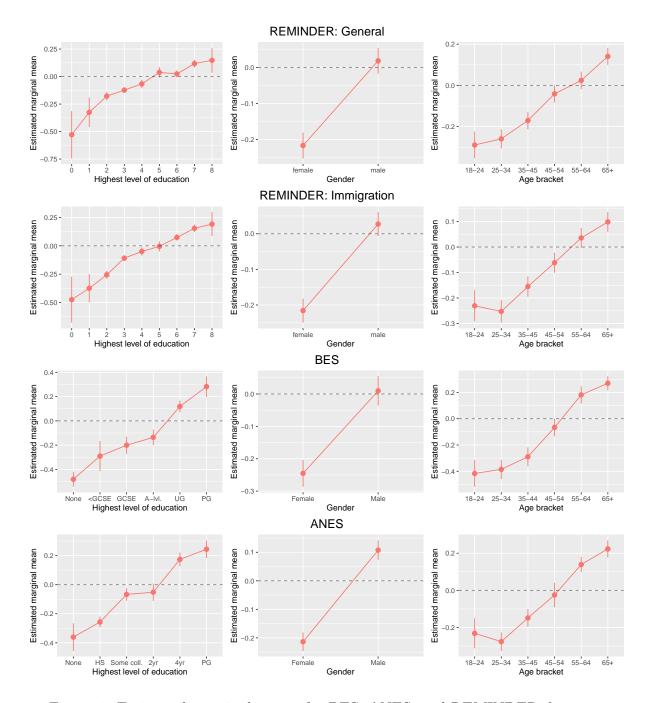


Figure 1: Estimated marginal means for BES, ANES, and REMINDER data sets

to knowledge about COVID in particular; and one relating to knowledge about climate change.

Our COVID data set comes from a a pre-registered survey experiment (N=2,917 UK citizens) fielded in the UK through the wake of its first wave in July 2020 (Allen and Ahlstrom-Vij, under review), containing the following knowledge items, which together formed a scale satisfying standard assumptions for IRT:

(COV1) COVID-19 can be transmitted in areas with hot and humid climate (true)

(COV2) There is currently no vaccine to protect against COVID-19 (true) (COV3) Most people who get COVID-19 recover from it (true)

The climate change data set is from the European Perceptions of Climate Change survey (Pidgeon 2016), which contains the following three knowledge items:

- (CC1) As far as you know, do you think the world's climate is changing or not? (Correct: Yes)
- (CC2) Thinking about the causes of climate change, which, if any, of the following best describes your opinion? (Correct: Climate change is partly, mainly, or completely caused by human activity)
- (CC3) To the best of your knowledge, what proportion of scientists agree that climate change is happening and that humans are largely causing it? (Correct: The vast majority of scientists agree, at 80% or more)

5 References

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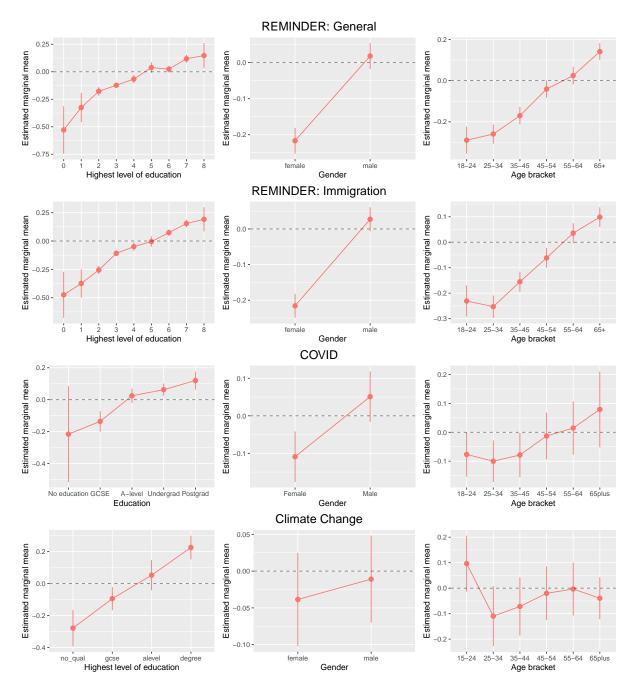


Figure 2: Estimated marginal means for REMINDER, COVID and Climate Change data sets $\,$