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People can understand IPCC visuals and are not influenced by colors

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20
21
22 8 **Abstract**

23 9 We carry out two online experiments with large representative sam-
24 10 ples of the U.S. population to study key climate visuals included in
25 11 the Sixth Report of the Intergovernmental Panel on Climate Change
26 12 (IPCC). In the first study (N=977), we test whether people can under-
27 13 stand such visuals, and we investigate whether color consistency within
28 14 and across visuals influences respondents' understanding, their atti-
29 15 tudes towards climate change and their policy preferences. Our findings
30 16 reveal that respondents exhibit a remarkably good understanding of
31 17 the IPCC visuals. Given that IPCC visuals convey complex multi-
32 18 layered information, our results suggest that the clarity of the visuals
33 19 is extremely high. Moreover, we observe that altering color consistency
34 20 has limited impact on the full sample of respondents, but affects the
35 21 understanding and the policy preferences of respondents who identify
36 22 as Republicans. In the second study (n=1169), we analyze the role
37 23 played by colors' semantic discriminability, that is the degree to which
38 24 observers can infer a unique mapping between the color and a concept
39 25 (for instance red and warmth have high semantic discriminability). We
40 26 observe that semantic discriminability does not affect attitudes towards
41 27 climate change or policy preferences and that increasing semantic
42 28 discriminability does not improve understanding of the climate visual.

43 29 **Keywords:** Climate change, IPCC Report, Climate visuals, Colors, Framing,
44 30 Visuals, Carbon Tax

1 Introduction

Between 2021 and 2022 the Intergovernmental Panel on Climate Change (IPCC) has released its Sixth Report (Tollefson et al, 2021). In light of the urgency of the threat posed by global warming, it is imperative that the message of the Report reaches a wide audience, as without broad support it will be hard to pass comprehensive and effective reforms addressing the climate crisis (Bernauer and McGrath, 2016; Ehret, 2021). Climate visuals can help ensuring that the IPCC message reaches the general public (Harold et al, 2016). First, visuals are an effective and efficient mean to convey scientific information (Fischer et al, 2018; Okan et al, 2012), as they harness the human visual system’s capacity to be a powerful pattern detector (Morelli et al, 2021; Franconeri et al, 2021). Second, many news media outlets included visuals from the Summary for Policymakers (SPM) of the IPCC Report in their articles (Table 1). In fact, “Visualizations have been the key element in the communication strategy of IPCC” (Xexakis and Trutnevyte, 2021).

Nevertheless, finding the right way to leverage the potential of data visualization is complex. A burgeoning literature is attempting to identify effective ways to convey climate-relevant information using visuals (Daron et al, 2015, 2021; O’Neill, 2017; Christel et al, 2018; Xexakis and Trutnevyte, 2021; Calvo et al, 2022), and some studies have focused specifically on visuals included in the IPCC reports (Taylor et al, 2015; Harold et al, 2020). Thus far, empirical evidence suggested that people have a limited understanding of IPCC visuals (Fischer et al, 2018; McMahon et al, 2015). However, these studies relied on small samples and focused on older versions of the IPCC Report.

Against this background, we carry out two between-subjects experiments with representative samples of approximately $N = 1000$ U.S. residents to investigate five research questions.

The first question we investigate is:

RQ1 *Do people understand the visuals included in the SPM of the Sixth IPCC Report? (Study I)*

To address this question, we test respondents’ understanding of two key IPCC visuals: SPM.3 and SPM.5(c).

We then turn to study the role of colors because IPCC experts have flagged colors as one of the key factors to guide the user in the experience of processing information (Morelli et al, 2021). To give an idea of how central the role of color is, the IPCC Visual Style Guide for Authors uses the word color 128 times in 28 pages. With respect to climate visualization, the research on colors has largely focused on how to convey uncertainty (Retchless and Brewer, 2016; Grigoryan and Rheingans, 2004; Viard et al, 2011) and how to identify the best color scale in quantitative mapping (Brewer et al, 1997; Harrower and Brewer, 2003; Dasgupta et al, 2018). We focus on different aspects and investigate the role played by consistency in color coding and semantic discriminability.

The IPCC aims for “consistent color coding” within and across reports (IPCC WGI Technical Support Unit, 2018). Consistency, however, can take

Newspaper	Country	Figure
Bloomberg	UK	SPM.3 and SPM.4
CNN	US	SPM.4
CNN	US	SPM.3
Eos	US	SPM.4
Forbes	US	SPM.4
Financial_Times	UK	SPM.4
Le_Monde	France	SPM.3
Straits_Times	Singapore	SPM.4
Sydney_Morning_Heralds	Australia	SPM.4
The_Economist	UK	SPM.3
Reuters	UK	SPM.5
Sky_News	UK	SPM.5
CBS_News	US	SPM.5

Table 1 A list of some of the news media that included in their articles the visuals from IPCC Report we use in Study I (SPM.3 and SPM.5) and Study II (SPM.4).

many forms. One way to apply consistent color coding is to always use the same color to describe an environmental event. Thus, for instance, one could always associate green to increases in precipitations. Another way is to always use the same color to describe events with a given connotation. Thus, for instance, one could always associate green to positive events and red to negative events. Figure SPM.3 prioritizes the first form of consistency (Figure 1, left panel).

SPM.3 is composed of three panels. The top panel describes observed changes in hot extremes, with increases marked in red and decrements marked in blue. The middle panel describes observed changes in heavy precipitations with increases marked in green and decrements marked in yellow. Last, the bottom panel describes observed changes in agricultural and ecological droughts with increases marked in yellow and decrements marked in green. This use of colors aims also at maximizing consistency across visuals, as other visuals in the Report use similar colors in association with these climate events¹. However, this use of colors creates an inconsistent association between colors and events' connotation. In the middle panel green denotes negative events, whereas in the bottom panel the same green denotes positive events.

In our treatment, we focus on the association between colors and events' connotation and mark in red all increases in extreme weather events because they all share a negative connotation, whereas we mark in green all decrements because they all share a positive connotation (Figure 1, right panel).

As this example shows, it is not always possible to simultaneously ensure consistency in all dimensions. Moreover, there might be trade-offs between preserving consistency within a visual and across visuals. From this perspective, testing understanding for SPM.3 and SPM.5(c) presents an important advantage. Even if our treatment improved within-visual consistency in terms of events' connotation, it reduced the consistency across visuals of the Report.

¹Note, however, that the colors used in the Figure SPM.3 are slightly different from both the colors used in the Visual Style Guide for Authors and the colors used in other graphs of the IPCC Report (e.g., SPM 5(c) and SPM 6 use different colors from SPM.3 to describe increases in precipitations).

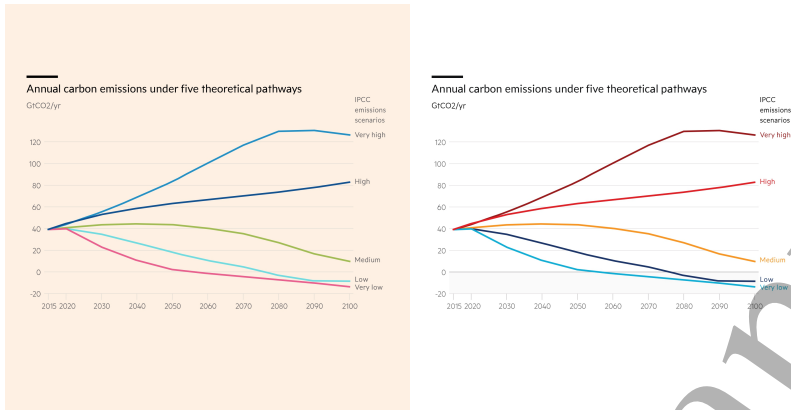


Fig. 2 The left panel represents SPM.4 as it appears in the Financial Times. The right panel represents our treatment, in which we used the colors of the IPCC Report.

panel of SPM.4 of the SPM describes five possible scenarios in terms of future CO_2 emissions (Figure 2, right panel). In this visual the curve describing the worst case scenario is in dark red, whereas the curve describing the best case scenario is in light blue. In one of its articles, the FT included a figure that is almost identical, but has curves of different colors (Figure 2, left panel). For instance, the curve describing the worst case scenario is in light blue, whereas the curve describing the best case scenario is pink. The colors used by the FT have a lower semantic discriminability, thus we investigate the following question:

RQ3 *Does using color with high semantic discriminability improve understanding? (Study II)*

Colors are not only important because they can aid or hinder understanding, but also because they can evoke emotions (Valdez and Mehrabian, 1994; Kaya and Epps, 2004). For instance, red is often associated with concepts like danger and fear (Pravossoudovitch et al, 2014; Jonauskaitė et al, 2019), whereas yellow is often connected with joy (Jonauskaitė et al, 2019). As colors affect emotions, they might shape the reaction to climate visuals, and in particular the level of concern for the climate crisis. Thus, the fourth question that we investigate is:

RQ4 *Do the colors used in a climate visual have an effect on respondents' level of concern for the climate crisis? (Studies I and II)*

Last, previous research has shown that some features of visuals can influence policy preferences (Romano et al, 2020). While colors should not be used to manipulate people's preferences, ensuring that the IPCC conveys information in a "policy-relevant but not policy-prescriptive" manner (Waisman et al, 2019) requires understanding the role played by colors. Thus, our fifth question is:

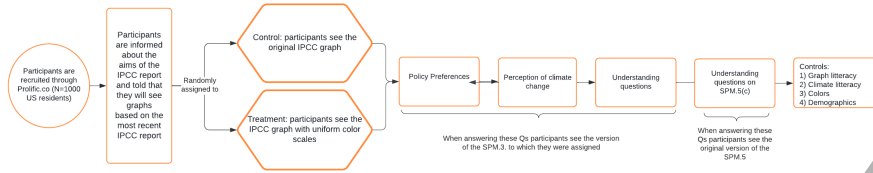
6 *Colors and Climate Visuals*

Fig. 3 Flow of Study I.

RQ5 Do the colors used in a climate visual affect respondents' policy preferences?
 (Studies I and II)

2 Materials and Methods

To answer our research questions, we carried out two large-scale experiments with representative samples of the U.S. population. There are at least two reasons to study whether the general public can understand IPCC visuals. First, key actors related with the IPCC explicitly stated that its Reports are also aimed to the general public (Lynn and Peeva, 2021). Second, many news media outlets relied on IPCC visuals (Table 1). Thus, visuals play a key role in ensuring that the IPCC messages reaches a wide audience.

Both experiments were pre-registered with AsPredicted (StudyI and StudyII).

Moreover, to obtain a precise measure of semantic discriminability we carried out two pre-surveys, which are described in details in the supplementary materials. The complete protocol of all surveys is available in the supplementary materials.

2.1 Study I

2.1.1 Visual and experimental manipulation

Our experimental manipulation in Study I relates to the main panel of the SPM.3 of the first part of the IPCC Report. The respondents who were randomly assigned to the control group saw SPM.3 as it appeared on the IPCC Report (Figure 1, left panel). Respondents who were randomly assigned to the treatment group saw the same visual, but in this case all the increases in extreme events were marked in red and the decreases in green (Figure 1, right panel). Figure 3 summarizes the flow of Study I.

2.1.2 Sample, Survey design and procedure

We sought to recruit a sample of $N = 1000$ U.S. residents on Prolific. Prolific provides researchers with the option to recruit a sample stratified across three demographics: age, sex and ethnicity. While our sample was representative across these dimensions, we note that Democrats were over-represented. This

is a standard feature of samples recruited online (Arechar and Rand, 2021). Respondents were paid \$1.1 (hourly compensation \$6.6).

At the beginning of the experiment participants saw a text containing information about the IPCC and the IPCC Report. They were also informed that the visuals they would see during the experiment were based on the IPCC Report. After the introduction, participants were randomly assigned to either the control group or the treatment group.

All participants were asked four sets of questions: (i) policy preferences (Table 2, top two rows); (ii) concerns for climate change (Table 2, bottom three rows); (iii) understanding of SPM.3 (Table 3); understanding of SPM.5(c) (Figure 4 and Table 4). Respondents answered questions *i*, *ii* and *iii* while seeing the version of SPM.3 to which they were assigned. They answered questions *iv* while seeing SPM.5(c). Before the understanding questions participants were asked to complete a simple attention check.

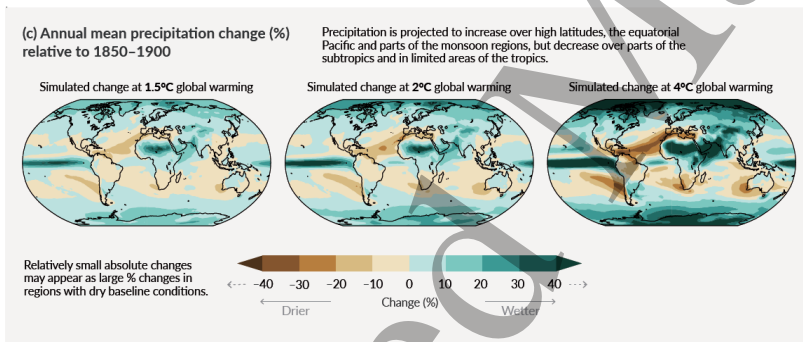


Fig. 4 Figure SPM.5(c). This figure was shown to respondents in Study I as it appears in the IPCC Report.

Understanding questions force respondents to think about the visual in a different way from which they would normally do when seeing the visual on a website. Therefore, they were included at the end to avoid anchoring the responses provided to the first two sets of questions.

The literature on graph comprehension identifies three levels of graph understanding (Friel et al, 2001; Galesic and Garcia-Retamero, 2011). The first level relates to the ability to read the data represented in the graph, e.g., by finding specific information. The second relates to the ability to identify relationship in the data. The third relates to the ability to extrapolate information from the data, e.g., by making predictions. As SPM.3 and SPM.5(c) do not convey information on trends over time, we investigate the first two dimensions of graph literacy: reading the data and identifying relationships in the data.

Afterwards, respondents answered questions that we use as control. The control questions can be grouped in: (i) graph literacy, (ii) climate literacy, (iii) color-related controls, (iv) standard demographic questions. Both the visual used to test graph literacy and the graph literacy questions were taken

8 *Colors and Climate Visuals*

Question	Range
Support for U.S. direct subsidies to the fossil fuel industry	0-100
Support for carbon tax	'strongly oppose' to 'strongly support'
How worried are you about global warming?	'not at all worried' to 'very worried'
How much do you think global warming will harm you personally? to 'strongly support'	'not at all' to 'a great deal'
When do you think global warming will start to harm people in the United States?	never' to 'They are being harmed right now'

Table 2 Questions used to assess participants' policy preferences and to study concerns for global warming. These questions were used in Study I and in Study II. The questions used to study concerns for global warming are taken from 'Climate Change in the American Mind: Beliefs & Attitudes'.

Question	Range	Understanding Level
Identifying the statements that correctly describes changes of extreme events in the area WNA (Western North America)? (SPM.3 Q1)	Five possible answers, of which one correct.	Level 1
Identifying if there areas in which agricultural and ecological droughts are decreasing but heavy precipitations are increasing (SPM.3 Q2)	Four answers, of which one correct	Level 1
Hot extremes have INCREASED in the MAJORITY of the areas (SPM.3 Q3)	True/False	Level 2
Heavy precipitations have INCREASED in the MAJORITY of the areas (SPM.3 Q4)	True/False	Level 2
'Agricultural and ecological drought have INCREASED in the MAJORITY of the areas' (SPM.3 Q5)	True/False	Level 1
'There are NO AREAS in which hot extremes have DECREASED' (SPM.3 Q6)	True/False	Level 1
'There are NO AREAS in which precipitations have DECREASED' (SPM.3 Q7)	True/False	Level 2
'There are NO AREAS in which Agricultural and ecological drought have DECREASED' (SPM.3 Q8)	True/False	Level 1
'There are more areas with medium confidence in the human contribution to changes in heavy precipitations than areas with medium confidence in the human contribution to changes in agricultural and ecological droughts' (SPM.3 Q9)	True/False	Level 2

Table 3 Questions used to assess participants' understanding of SPM.3. These questions were used only in Study I.

Question	Range	Understanding Level
Please look at the graph above. Are there more areas in which precipitations decrease by 30% or more in the 1.5°C (34.7°F) global warming scenario or in the 4°C (39.2°F) global warming scenario? (SPM.5(c) Q1)	Four possible answers, of which one correct.	Level 2
Please look at the graph above. Are there more areas in which precipitations increase by 30% or more in the 1.5°C (34.7°F) global warming scenario or in the 4°C (39.2°F) global warming scenario? (SPM.5(c) Q1)	Four answers, of which one correct	Level 2

Table 4 Questions used to assess participants' understanding of SPM.5(c). These questions were used only in Study I.

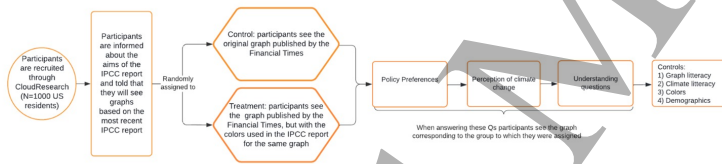


Fig. 5 Flow of Study II.

from Galesic & Garcia-Retamero (2011), whereas climate literacy questions have been adapted from Leiserowitz et al. (2011).

2.2 Study II

2.2.1 Visual and experimental manipulation

Our experimental manipulation in Study II relates to the main panel of SPM.4. SPM.4 was included by the FT in one of its articles, but the FT used different colors from those used by the IPCC. The participants who were randomly assigned to the control group saw SPM.4 as it appeared on the Financial Times (Figure 2, left panel). Participants who were randomly assigned to the control group saw the same visual, but with the colors used in SPM.4 (Figure 2, right panel). Figure 5 summarizes the flow of Study II.

2.2.2 Sample, Survey design and procedure

We sought to recruit a representative sample of $N = 1000$ U.S. residents on Cloudresearch. Unlike Prolific, Cloudresearch does not automatically provide a representative sample. Thus, to ensure that our sample was stratified across the same demographics we launched the experiment several times, creating restrictions by age, gender and race matching those we obtained in Study I on Prolific. As soon as the target quotas were reached, the experiment was

Question	Range	Understanding Level
Estimating when in the “very high” scenario $GtCO_2/yr$ will reach 100? (SPM.4 Q1)	Five possible answers, of which one correct.	Level 1
Estimating distance at various points in time between curves representing various scenarios (SPM.4 Q2)	Respondents must rank the possible alternatives	Level 2
Formulating predictions based on the scenarios (SPM.4 Q3)	Four possible answers, one of which correct.	Level 3

Table 5 Questions used to assess participants’ understanding of SPM.4. These questions were used only in Study II.

closed for that category of respondents. Respondents were paid \$0.9 (hourly compensation \$6.75). At the beginning of the experiment, all respondents saw the same message shown in study one. After the introduction, participants were randomly assigned to either the control group or the treatment group. All participants were then asked three sets of questions: (i) policy preferences; (ii) concerns for climate change (iii); understanding of SPM.4.

The first two groups of questions were the same as in Study I (Tables 2). The understanding questions are reported in Table 5. Respondents answered all these questions while seeing the figure to which they were assigned. After answering the understanding questions, respondents answered the same control questions as in Study I. Respondents were also asked to complete the same attention check as in Study I.

3 Results

Tables 6 and 7 report respectively the summary statistics for Study I and Study II.

3.1 Understanding of the IPCC visuals (Study I)

First, we analyze the level of understanding of the original SPM.3 and SPM.5(c). Figure 6 left panel indicates the percentage of correct answers provided by respondents to the understating questions related to SPM.3 and SPM.5(c) disaggregated by level of understanding. We observe that for the levels of understanding tested the percentage of correct answers is close to 70% for both visuals. The right panel of Figure 6 shows that reducing consistency between SPM.3 and SPM.5(c) did not worsen the understanding of SPM.5(c), the second visual seen by respondents.

3.2 Consistency of colors in climate visuals (Study I)

We then compare the understanding of the control group, who saw SPM.3 with original colors (Figure 1 left panel), and of the treatment group, who saw our version of SPM.3 (Figure 1 right panel). We observe no significant differences

	(All)		(Control)		(Treatment)	
	mean	sd	mean	sd	mean	sd
Preferred subsidy (in billions)	18.14	18.54	17.32	17.64	18.90	19.32
Support for a carbon tax	2.15	1.47	2.15	1.48	2.15	1.46
Worry about climate change	2.94	1.25	2.96	1.24	2.92	1.26
Global warming: personal harm	1.76	0.93	1.76	0.92	1.75	0.93
Global warming: when does it hurt the US?	3.89	1.61	3.93	1.57	3.85	1.64
Correct answer SPM.3 Q1 (%)	0.63	0.48	0.64	0.48	0.63	0.48
Correct answer SPM.3 Q2 (%)	0.63	0.48	0.60	0.49	0.66	0.47
Correct answer SPM.3 Q3 (%)	0.94	0.24	0.95	0.23	0.94	0.24
Correct answer SPM.3 Q4 (%)	0.49	0.50	0.49	0.50	0.50	0.50
Correct answer SPM.3 Q5 (%)	0.59	0.49	0.56	0.50	0.61	0.49
Correct answer SPM.3 Q6 (%)	0.84	0.37	0.84	0.37	0.84	0.37
Correct answer SPM.3 Q7 (%)	0.76	0.43	0.73	0.44	0.79	0.41
Correct answer SPM.3 Q8 (%)	0.87	0.34	0.87	0.33	0.86	0.34
Correct answer SPM.3 Q9 (%)	0.53	0.50	0.54	0.50	0.52	0.50
Tot. understanding SPM.3 (score out of 9)	6.29	2.09	6.22	2.02	6.35	2.15
Correct answer SPM.5(c) Q1 (%)	0.68	0.47	0.66	0.48	0.70	0.46
Correct answer SPM.5(c) Q2 (%)	0.71	0.45	0.71	0.46	0.72	0.45
Tot. understanding SPM.5(c) (%)	0.70	0.38	0.68	0.38	0.71	0.38
Observations	977		476		501	

Table 6 Summary statistics (mean and standard deviation) for Study I for the full sample and by treatment group.

	(1)		(2)		(3)	
	mean	sd	mean	sd	mean	sd
Preferred subsidy (in billions)	22.25	20.86	21.57	20.50	22.92	21.22
Support for a carbon tax	1.96	1.45	1.95	1.45	1.97	1.45
Worry about climate change	2.79	1.32	2.85	1.30	2.72	1.33
Global warming: personal harm	1.67	0.96	1.70	0.94	1.64	0.97
Global warming: when does it hurt the US?	3.64	1.79	3.72	1.76	3.55	1.82
Correct answer SPM.4 Q1 (%)	0.79	0.40	0.78	0.41	0.80	0.40
Correct answer SPM.4 Q2 (score out of 4)	1.25	1.41	1.23	1.40	1.26	1.42
Correct answer SPM.4 Q3 (%)	0.40	0.49	0.40	0.49	0.41	0.49
Total understanding SPM.4 (score out of 6)	2.44	1.76	2.41	1.75	2.48	1.77
Observations	1169		586		583	

Table 7 Summary statistics (mean and standard deviation) for Study II for the full sample and by treatment group.

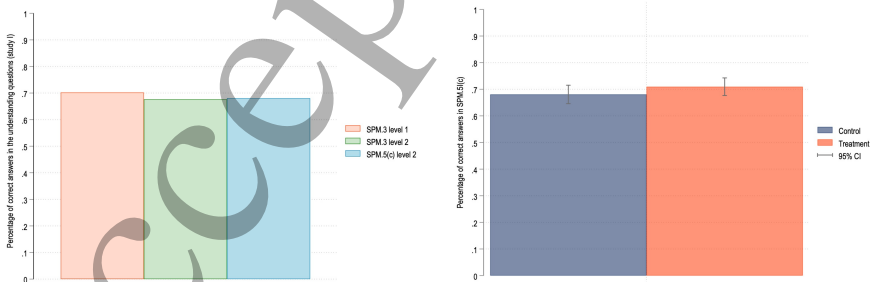


Fig. 6 The left panel shows the percentage of participants in Study I answering correctly questions aimed at testing Level-1 and Level-2 understanding of SPM.3 and Level-2 understanding of SPM.5(c). The right panel shows the percentage of participants in Study I answering correctly to understanding questions related to SPM.5(c) divided by treatment and control.

1
2 256 between the treatment and the control group. The full regression tables are
3 257 available in the supplementary materials.

5 258 **3.3 Semantic discriminability (Study II)**

6 259 We now turn to our third research question: whether using colors that have a
7 260 higher semantic discriminability improves understanding of the climate visual
8 261 as hypothesized by the literature (Terrado et al, 2022). We find no evidence
9 262 that using colors with higher semantic discriminability improves understanding.
10 263 We observe no significant difference in the understanding of the visual
11 264 between the control and the treatment group. The full regression tables are
12 265 available in the supplementary materials.

14 266 **3.4 Colors, concern for the climate crisis and impact on** 15 267 **policy preferences (Studies I and II)**

16 268 We then analyze whether altering colors in climate visuals affects how concerned
17 269 people are about global warming. In both studies, we observe no effect of
18 270 colors on concerns for global warming. Similarly, we find no impact of colors on
19 271 policy preferences. The full regression tables are available in the supplementary
20 272 materials.

22 273 **3.5 Heterogeneous treatment effects**

23 274 Last, we look at heterogeneous treatment effects for age (young-adult-older),
24 275 gender (female-male-other) and political affiliation (democrat-republican-no
25 276 strong preference). We observe no results when looking at age and gender.
26 277 Instead, we observe interesting results when focusing on political affiliation.

27 278 First, in Study I respondents who identify as Republicans who saw the
28 279 modified version of SPM.3 (treatment group) display a better overall under-
29 280 standing ($p = 0.002$ in the model with all controls). These results are robust
30 281 to different sets of controls (see supplementary materials). We do not find sim-
31 282 ilar results for Democrats or respondents who state that they have no strong
32 283 preference for either party.

33 284 Moreover, we find that both in Study I and II the treatments affect
34 285 Republicans' stated policy preferences (Table 8).

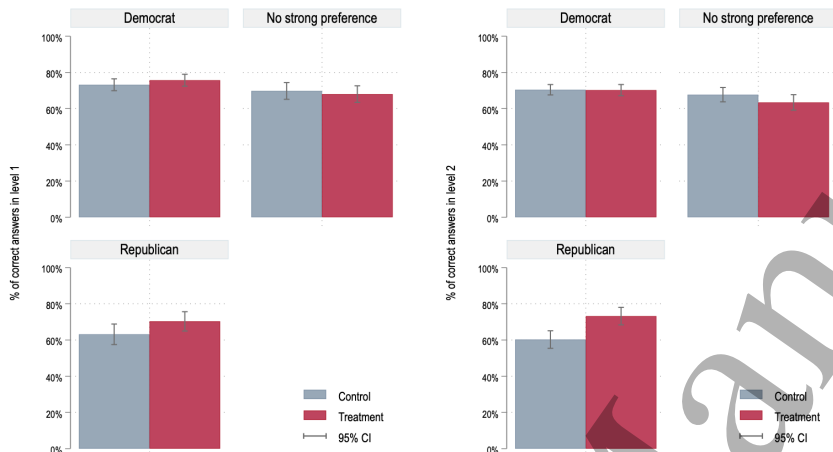


Fig. 7 Percentage of respondents who provided the correct answer to the understanding of SPM.3 by level of understanding. The results are presented for the full sample and disaggregated by political affiliation.

Policy	Rep	Dems	No strong preference	TE Reps	TE Others
Study I: subsidy	25.1 (18.52)	13.3 (15.4)	22.1 (21.11)	6.83** p=0.02	1.91 p=0.5
Study I: tax	1.1 (1.2)	2.7 (1.27)	1.8 (1.47)	0.899*** p=0.008	-0.12 p=0.72
Study II: subsidy	27.5 (20.81)	18.6 (19.9)	23.3 (21.27)	5.91* p=0.052	- 5.21 p=0.101
Study II: tax	0.86 (1.21)	2.63 (1.2)	1.88 (1.39)	-0.27 p=0.43	0.5 p=0.12

Table 8 The impact of the treatments on policy preferences. The table reports the results of regressions with a fully interacted model accounting for the different impact of the treatment on respondents with different political affiliations in Study I (top two rows) and Study II (bottom two rows). The specification includes a dummy equal to one for participants in the treatment, a dummy equal to one for participants identifying as republicans, a dummy equal to one for participants not identifying as democrats or republicans and interactions terms for republicans in the treatment group and respondents with no strong preference for either party in the treatment. The first three columns report the mean and standard deviation (in parentheses) for the variable considered for Republicans (Rep), Democrats (Dems) and No strong preference. Columns 4 and 5 report the coefficient of the interaction term between the treatment and that political affiliation, along with the p-value obtained from a regression controlling for the treatment, the interactions and controls (demographics, the level of worry about global warming, the level of understanding of the figure, the level of graph literacy, the level of climate literacy and participants' feelings about the color scale used). We study the impact of the treatment on the desired level of subsidies using an OLS regression with robust standard errors. We study the impact of the treatment on the support for a carbon tax through an ordered logit model with robust standard errors. Full regression tables and robustness checks can be found in the supplementary materials.

1
2 286 In particular, in Study I we observe that respondents who identify as
3 287 Republicans included in the treatment state that they want higher direct sub-
4 288 sidies for fossil fuels ($p = 0.02$ in the model with all controls), while also
5 289 indicating a higher support for a carbon tax ($p = 0.008$ in the model with all
6 290 controls). Moreover, we observe that also in Study II respondents who identify
7 291 as Republicans included in the treatment support higher subsidies for fossil
8 292 fuels ($p = 0.052$ in the model with all controls). We do not find similar results
9 293 for Democrats or respondents who state that they have no strong preference
10 294 for either party.

11 12 295 **4 Discussion**

13 14 296 **4.1 Do people understand the visuals included in the** 15 297 **SPM of the IPCC Report?**

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17 298 Scholars who carried out studies analyzing people's understanding of visuals
18 299 included in previous versions of the IPCC Report had lamented that the results
19 300 were disappointing (McMahon et al, 2015; Fischer et al, 2020). On the contrary,
20 301 we report that respondents show a remarkably high level of understanding of
21 302 the visuals we analyze.

22 303 While there are no universally accepted thresholds to determine whether
23 304 "enough" people have understood a visual, two comparisons help putting these
24 305 results in perspective. First, the percentage of correct answers provided to
25 306 understanding questions of IPCC visuals is close to the percentage of correct
26 307 answers provided to the questions used to test graph literacy. But the IPCC
27 308 visuals conveyed complex multi-layered information, whereas the graph liter-
28 309 acy visual merely represented a linear trend. Second, the percentage of correct
29 310 responses provided to our understanding questions is higher than that provided
30 311 by experts in previous research focusing on older versions of the IPCC Report
31 312 (Fischer et al, 2020). Thus, a sample of experts might display an even better
32 313 understanding of the visuals used in the Sixth IPCC Report. Importantly, we
33 314 note that respondents had limited incentives to understand the visuals because
34 315 their payment was not conditional on them providing the right answer.

35 316 These considerations suggest that the authors of the visuals of the SPM of
36 317 the IPCC Report did a remarkable job in conveying complex information in a
37 318 clear and understandable manner.

38 39 319 **4.2 Consistent color coding and understanding**

40 320 Using consistent color coding is among the recommendations given in the IPCC
41 321 Visual Style Guide. However, there is no univocal way to define consistency.
42 322 In Study I we compared two ways to interpret the idea of using consistency.
43 323 In the framing used by the SPM.3, the colors were consistently associated
44 324 with a given environmental event. Thus, for instance, increases in precipitation
45 325 were always marked in green. In our treatment, the colors were consistently

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2 326 associated with a given connotation of an event. Consequently, all positive
3 327 events were marked in green and all negative events were marked in red.

4 328 We observe that the visual shown in the treatment was better understood
5 329 by Republicans than SPM.3. We did not observe the same effect for Democrats,
6 330 possibly because they displayed a remarkably high level of understanding
7 331 already in the control group. Therefore, there might have been a ceiling effect.
8 332 Moreover, we observed that reducing color consistency across visuals did not
9 333 reduce understanding of the second visual. This suggests that consistent color
10 334 coding within a given visual is more important than consistent color coding
11 335 across visuals. This is even truer, because many news media included visuals
12 336 from the SPM in their articles, and generally they did not include all IPCC
13 337 visuals in the same article. And since people are likely to learn about the IPCC
14 338 Report from media, there are practical reasons to prioritize color consistency
15 339 within visuals over consistency across visuals.

16 340 One possibility is that the better understanding achieved by the treatment
17 341 group is driven by the higher semantic discriminability of the colors used in
18 342 our treatment. In fact, green is often associated with gains, whereas red is
19 343 associated with losses (Fischer et al, 2020). Instead, in the original visual of
20 344 the IPCC the color green was mostly used to indicate negative event. While
21 345 we cannot rule out this alternative explanation, we note that in Study II we
22 346 find no impact of semantic discriminability.

23 24 347 **4.3 Semantic discriminability and understanding**

25 348 In Study II we analyzed the role of semantic discriminability. The control group
26 349 saw the visual with the low discriminability colors used by the FT, whereas the
27 350 treatment group saw the visual with the high discriminability colors used by
28 351 the IPCC. In line with the literature (Terrado et al, 2022), we were expecting
29 352 high semantic discriminability to foster understanding. However, we observed
30 353 no significant differences between the treatment and the control group.

31 354 Clearly, our results do not imply that semantic discriminability is never
32 355 important. However, they show that the relationship between semantic dis-
33 356 criminability and understanding is likely to be nuanced and context dependent,
34 357 instead of monotonic and universal.

36 37 358 **4.4 Colors and concern for climate change**

38 359 Colors are known to have an impact on emotions (Jonaskaite et al, 2019).
39 360 Thus, an important question is whether the colors of a visual might affect
40 361 the emotional response to the visual. For example, using colors associated
41 362 with calm and peacefulness might lead people to underestimate the dangers
42 363 conveyed by a climate visual. Contrarily to our expectations we observe that
43 364 colors have no impact on how concerned people are about the climate crisis.
44 365 One possible explanation for this finding is that climate change is a highly
45 366 polarized and debated topic, and hence people are likely to have been exposed
46 367 to information and partisan cues (Goldberg et al, 2021). To put it differently,
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they are “pre-treated” by media (Bernauer and McGrath, 2016). Nevertheless, we do not suggest that colors never have an impact on the how concerned people are about the climate crisis.

4.5 Colors and policy preferences

Recent evidence suggests that the features of a visual can affect policy preferences (Romano et al, 2020). However, we find that colors affect policy preferences only of respondents who identify as Republicans (Table 8). Our results are consistent with recent evidence suggesting that conservatives’ preferences with respect to climate policies are affected by framing (Marlow and Makovi, 2023). Intriguingly, in Study I we observe that Republicans in the treatment group state higher preferred subsidies for fossil fuels but they also indicated a higher support for a carbon tax. These are not the results that we were expecting, therefore we do not advance a post-hoc hypothesis. However, we emphasize that our results provide strong preliminary evidence that colors of climate visuals affect policy preferences. Consequently, we believe that more studies are needed to understand through which mechanisms colors influence policy preferences.

5 Conclusions

Empirical evidence suggested that the visuals included in the previous versions of the IPCC Report were often misinterpreted. We run two large scale surveys with two representative samples of the U.S. population and find that people show a remarkably good understanding of the visuals used in the most recent version of the IPCC Report. Moreover, we investigated the role that colors play in climate visuals and the extent to which they affect understanding of the visuals, policy preferences and concerns for the climate crisis. This study only focused on three visuals from the IPCC Report and only included U.S. respondents. Future studies should test whether also other visuals are equally clear, and if also people from different countries display a good level of understanding.

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7 Competing interests

The authors declare no competing interests.

8 Ethical Statement

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9 Data Availability

The data are fully available upon request.