

1 What is policy analytics? An exploration of 5 years of environmental management applications

2

3 Yves Meinard ^{1,#}, Olivier Barreteau ², Christophe Boschet ³, Katherine A. Daniell ^{4,5,6}, Nils
4 Ferrand ², Sabine Girard ⁷, Joseph H.A. Guillaume ⁴, Emeline Hassenforder ², Matthew Lord ⁴,
5 Myriam Merad ¹, Ehsan Nabavi ^{6,8}, Claire Petitjean ⁹, Irene Pluchinotta ¹⁰, Juliette Rouchier ¹,
6 Alexis Tsoukias ¹, Pascale Zarate ¹¹

7

8 ¹ Université Paris-Dauphine, PSL Research University, CNRS, UMR [7243], LAMSADE, F-
9 75016 Paris, France

10 ² G-EAU, AgroParisTech, Cirad, IRD, INRAE, Montpellier SupAgro, Univ Montpellier,
11 Montpellier, France

12 ³ INRAE, UR ETBX, 33612 Cestas, France

13 ⁴ Australian National University, Fenner School of Environment & Society, Canberra, Australia

14 ⁵ Australian National University, 3A Institute, College of Engineering & Computer Science,
15 Canberra, Australia

16 ⁶ Australian National University, Centre for European Studies, School of Politics and
17 International Relations, Canberra, Australia

18 ⁷ INRAE, UR LESSEM, INRAE, Grenoble

19 ⁸ Australian National University, Centre for the Public Awareness of Science, Canberra,
20 Australia

21 ⁹ Syndicat Mixte de la Rivière Drôme, Saillans, France

22 ¹⁰ UCL Institute for Environmental Design and Engineering, The Bartlett Faculty of the Built
23 Environment, University College London, UK

24 ¹¹ University of Toulouse, UMR 5505 IRIT, Toulouse, France

25

26 # Corresponding author: yves.meinard@lamsade.dauphine.fr

27

28 Abstract

29 Our digital age is characterized by both a generalized access to data and an increased call for
30 participation of the public and other stakeholders and communities in policy design and decision-
31 making. This context raises new challenges for political decision-makers and analysts in providing
32 these actors with new means and moral duties for decision support, including in the area of
33 environmental policy. The concept of “policy analytics” was introduced in 2013 as an attempt to
34 develop a framework, tools and methods to address these challenges. This conceptual initiative
35 prompted numerous research teams to develop empirical applications of this framework and to
36 reflect on their own decision-support practice at the science-policy interface in various
37 environmental domains around the world. During a workshop in Paris in 2018, participants shared

38 and discussed their experiences of these applications and practices. In this article we present and
39 analyze a set of applications to identify a series of key properties that underpin a policy analytics
40 approach, in order to provide the conceptual foundation for policy analytics to address current
41 policy design and decision-making challenges. The induced properties are demand orientedness,
42 performativity, normative transparency and data meaningfulness. We show how these properties
43 materialized through these six case-studies, and we explain why we consider them key to effective
44 policy analytics applications, particularly in environmental policy design and decision making on
45 environmental issues. This clarification of the policy analytics concept eventually enables us to
46 highlight research frontiers to further improve the concept.

47 Keywords: decision support; environmental policies; legitimacy; data; policy analytics

48

49

50 1. Introduction

51 The digital age has provided access to multiple sources of data and information for an increasing
52 part of the world's population and has accelerated opportunities for their analysis, including
53 through increased computational capacity. At the same time, the demand for opening policy-
54 making processes to stakeholders, communities and the general public has evolved into a
55 generalized call for more inclusive and extensive participation, in some cases becoming entrenched
56 in national or supra-national regulations. This has often generated conflicting understandings of
57 problems, driven by multiple bodies of expertise and knowledge on the same issues, which are
58 embodied by diverse actors in society (see for example Arts et al. 2017). Since the expansion of
59 environmental movements in the 1970s and 1980s around conservation and environmental

60 protection, the environmental policy domain has long been a prominent arena for the tension
61 between these two trends (increased information availability and calls for participation) (e.g.
62 O'Donnell et al., 2019; Long, 2019). However, the current digital age has rapidly exacerbated the
63 availability of multiple, and at times contradictory, bodies of information.

64 This context raises new challenges and opportunities for innovatively engaging citizens in
65 decision-making, and improving policy makers' capacities to intervene effectively in complex
66 problems. In recent years, Government actors have more actively sought to address both the
67 opportunities and challenges of new demands and capabilities driven by technological change, as
68 highlighted by the proliferation of various dedicated policy and legislative instruments, such as the
69 General Data Privacy Regulation (GDPR) in Europe, and high-level strategies developed by the
70 US, China, France, Germany, and Australia (e.g. DISS 2018, Federal Data Strategy 2019, FMEAE
71 2018, The White House 2019, Villani 2018, Webster et al. 2019).

72 Parallel to, and in support of these shifts, academic research is also seeking to formalize new
73 models of decision support to environmental policies, to enable a productive interplay between the
74 use of new information technologies and the enhanced public participation. Among these
75 initiatives, policy analytics, as formalized in Tsoukias et al. (2013) and Daniell et al. (2015),
76 provides a framework, tools and methods fit for purpose. The term 'analytics', has historically
77 been used for decision support within individual sectors, with previous research focusing on areas
78 such as 'business analytics', 'health analytics' and 'learning analytics'. Across these applications,
79 the term 'analytics' is understood as an umbrella term describing a variety of analytical methods
80 and approaches with a sophistication that can match the complexity of the data types (both
81 qualitative and quantitative), processing and analysis demands of the digital age (Tsoukias *et al.*,
82 2013). Tsoukias *et al.* (2013) wanted to promote the use of such 'analytics' tools to address the

83 public policy issues for which they may be relevant. However, Tsoukias *et al.* (2013) also stressed
84 the relative difficulty of applying ‘analytics’ within the public realm, mainly due to the unique
85 constraints associated with decision support of public policies; in particular, the use of public
86 money and the associated need for transparency, the prevalence of participatory and deliberative
87 processes, and the non-monetary and multifaceted nature of policy goals. To capture this two-fold
88 ambition, they defined “policy analytics” as a project to “*support policy makers in a way that is*
89 *meaningful (in a sense of being relevant and adding value to the process), operational (in a sense*
90 *of being practically feasible) and legitimating (in the sense of ensuring transparency and*
91 *accountability), [by drawing] on a wide range of existing data and knowledge (including factual*
92 *information, scientific knowledge, and expert knowledge in its many forms) and [combining] this*
93 *with a constructive approach to surfacing, modelling and understanding the opinions, values and*
94 *judgements of the range of relevant stakeholders”.*

95 This concept of “policy analytics” has aroused interest among many researchers in the
96 environmental policy domain in recent years, with numerous discussions about its utility and
97 possible improvements, and several applications in the field being held in different places around
98 the world. This article aims to draw on these discussions and applications to clarify the policy
99 analytics concept so that its use and relevance can be clarified and expanded. To that end, we
100 analyze a series of examples of concrete applications of the policy analytics framework to
101 environmental policies. We first outline our methodological approach for clarifying the concept
102 (section 2). We then implement this approach (section 3). We present our series of case studies
103 (subsection 3.1). We then articulate four normative properties that emerged from the discussions
104 and comparisons of these case studies (subsection 3.2). These properties constitute the core of our

105 proposed improved definition of policy analytics. Lastly, section 4 outlines avenues for future
106 research on and around policy analytics.

107

108 2. A methodology to rethink “policy analytics” as an approach to support environmental
109 decision makers

110 In their context of launching a research dynamic, Tsoukias et al. (2013) proposed a deliberately
111 wide definition of policy analytics in order to encourage discussions with a diverse and inter-
112 disciplinary group of researchers, policy officials and data industry collaborators. This strategy
113 proved effective, and a series of research projects were launched and developed, as part of an effort
114 to develop and gain traction for the policy analytics concept and its application. However, this type
115 of approach, which uses a more general definition to avoid excluding useful contributions, also
116 has its limits, especially once the concept is mature enough to be compared with alternative
117 frameworks.

118 As it happens, numerous other frameworks also attempt to address the challenges associated with
119 developing public policy in a highly data driven age, including “policy informatics” (Johnston,
120 2015), “computational social sciences” (Lazer et al., 2009), “big data in public affairs” (Mergel et
121 al., 2016), and “utilization-focused” and “systemic evaluation” of public policies (Midley 2006;
122 Boyd et al. 2007; Patton 2008). Shared amongst these various frameworks is the acknowledgement
123 that our current information, communication and technological environment is undergoing rapid
124 changes, and consequently there is both a need and an opportunity for public policy to utilize the
125 capabilities of changing information and communication technologies. Furthermore, these
126 approaches also agree on the issues that will emerge from increased usage of data in both public

127 and private settings, including questions around privacy, legitimacy, and accountability, and the
128 need for new regulatory approaches that mandate certain standards in relation to these governance
129 attributes.

130 As various research teams began to attempt real-world applications of the policy analytics concept,
131 the lack of specificity in the definition prompted discussions on the definition itself, and on what
132 made policy analytics unique from the alternative frameworks highlighted earlier. Various papers
133 have proposed alternative definitions based on proposed clarifications of one or several of the
134 criteria mentioned in Tsoukias et al. (2013). Jeanmougin et al. (2017) proposed to formalize
135 Tsoukias et al. (2013)'s definition, using policy analytics as an evaluation framework applied to a
136 conservation policy, by singling out four elementary criteria, associated with concrete examples.

137 As compared with Tsoukias et al. (2013), this formulation retains the operationality and legitimacy
138 criteria, but replaces the "meaningfulness" requirement, which they considered to be too vague,
139 by two criteria referring, respectively, to a "scientificity" requirement and a requirement to bring
140 in a demonstrable contribution. However, this clarification focused on a specific usage of the
141 policy analytics framework (as an evaluation tool) and applied to a specific context (i.e.
142 conservation policies). Jeanmougin et al. (2017) also highlighted the difficulty substantiating the
143 "legitimacy" requirement at the core of the policy analytics framework. Meinard (2017) attempted
144 to clarify this requirement by proposing an open-ended list of legitimacy criteria, but here again
145 this attempt was focused on the specific context of conservation policies. Interestingly, some of
146 the criteria proposed referred to the scientific credentials of the policies whose legitimacy was
147 being evaluated, highlighting that the four criteria proposed by Jeanmougin et al. (2017) are not
148 completely independent. Although this interdependency between some of the criteria constituting
149 the definition is not necessarily a fatal flaw, a definition based on independent criteria would

150 certainly be clearer. In the same vein, Choulak et al. (2019) briefly discussed the vagueness of the
151 operationality criterion.

152 The need to clarify the definition and the risks associated with too rigid definitions were discussed
153 in numerous internal seminars among researchers in the group, based on applications of various
154 versions of the framework based on a broader variety of policy issues, including the above
155 mentioned environmental issues but also public health problems (Richard et al. 2018) and public
156 management issues (Touret et al. 2019). In the wake of theoretical work clarifying the difference
157 between tools, methods and approaches in decision support theories and practices (Meinard &
158 Tsoukias 2019), these discussions pointed to the conclusion that policy analytics is neither a field
159 (such as, for example, policy analysis) nor a tool nor a methodology (such as, for example, focus
160 groups or other participatory tools), but rather an “approach” to decision support intended for
161 actors in public policy decision making. Following Meinard & Tsoukias (2019), we use the term
162 “approach” here to refer to “a way by which [an analyst] conducts a [decision support] process”.
163 A given approach can be applied to different issues, which can belong to different academic fields,
164 and it can make use of a variety of methodologies, which can themselves be used by different
165 approaches. In this understanding, which is anchored in Habermas’s epistemological views
166 (Habermas 1985, 1990), “approaches” are defined by normative properties that specify key aspects
167 of the way analysts should use available tools and methods.

168 This view of policy analytics as an approach embodying normative properties opens avenues to
169 complement the top-down definitional approach used in these previous works by identifying,
170 through a bottom-up procedure, normative properties, to some extent shared by exemplary case
171 studies, which could be considered to provide an addition to the definition of policy analytics.
172 Because the case studies explored below were performed with policy analytics in mind, they can

173 be seen as partial but complementary attempts to clarify an underlying ambition shared by all the
174 researchers who decided to gather under the banner of “policy analytics”.

175 In this dynamic, during a workshop in Paris in 2018, a series of examples of policy analytics
176 applications to environmental policies have been shared and discussed by participants. These
177 applications provided the empirical material to venture a formulation of key properties, in an
178 abductive approach (Peirce 1966). This formulation was then used in a reconstructive approach to
179 rationalize some key aspects of the applications. The results of this reconstruction are presented in
180 the next section.

181 We should emphasize at the outset that efforts to clarify the definition in this way are not doomed
182 to constrain the potential of the concept, as Tsoukias et al. (2013) feared. As long as the definition
183 remains open-ended and open to discussion and improvements, attempts to refine it can usefully
184 clarify the underlying ambitions of different policy analytics research programs and provide
185 directions for future investigations.

186

187 3. Conceptualizing policy analytics: lessons from 5 years of applications

188 Using the methodology delineated above, in the present section, we start by describing the 6 case
189 studies that were discussed in the 2018 workshop (3.1). The descriptions are all organized in the
190 same way: We start by explaining the context (what is the policy at issue, what are the processes
191 engaged) (1). We then explain the reasons why the researchers involved conceived of their works
192 as applications of the policy analytics concept. Because, as explained in previous section, the
193 original definition of policy analytics was quite open, these reasons were disparate and, very often,
194 focused on quite different interpretations of the concept (2). We then describe the data produced

195 and/or analyzed (3). We finish by summarizing the outcome of each policy analytics application
196 (4).

197 Following this description of the case studies, we articulate the four normative properties that
198 emerged from the discussions and comparisons of case studies, which we propose as candidates to
199 structure an improved definition of policy analytics (3.2).

200

201 3.1. Examples of applications

202 3.1.1. Case 1: Elaboration of a wetland prioritization platform

203 (1) The first case involved the elaboration of an operational wetland prioritization platform in
204 Bourgogne-France-Comté (Choulak et al. 2019) that would be seen as legitimate by its key
205 stakeholders. Wetlands are ecosystems whose functioning is largely determined by water,
206 such as swamps, alluvial forests, bogs, etc. These ecosystems are the target of numerous
207 conservation policies around the world, including the RAMSAR convention, and dedicated
208 legislation in France. Wetland prioritization is a crucial step in most action plans devoted to
209 conserving or restoring wetlands in line with these policies. It consists of using available data
210 on wetlands (e.g. ecological features, hydraulic functions) and the context (e.g. urbanization
211 dynamics, land use) to decide on which wetlands managers should prioritize. In 2017, the
212 “wetland taskforce” (“Pôle Milieux Humides”) of the Bourgogne-Franche-Comté region
213 (France)—a team within a non-profit environmental organization (Conservatoire Espaces
214 Naturels)—was entrusted to elaborate a spatialized database on wetlands by a consortium of
215 regional to national scale institutions funding environmental actions. It was to focus on the
216 whole regional scale based on a new prioritization methodology that would also need to be
217 elaborated.

218 (2) Relevant databases available for prioritizing wetlands are large and heterogeneous, and very
219 often standard practices tend to conflate very different kinds of data indiscriminately. Some
220 of the databases house quantitative scientific data such as the results of hydrological models
221 or data on the abundance of a given species. Others have political aspects and may include
222 different forms of qualitative and quantitative information, such as zoning maps produced
223 through political processes. Tsoukias et al. (2013) emphasized the importance of taking into
224 account the nature and design of data to provide relevant and legitimate decision support. The
225 researchers involved in this case study therefore saw standard practices in wetland
226 prioritization as an example domain in which policy analytics could make a difference, by
227 developing methods that give importance to the nature of the data they use and their design.

228 (3) The data used were the contents of the spatialized database elaborated by the wetland taskforce,
229 and all the metadata corresponding to the methodologies used to capture these data, which we
230 used to develop rules to aggregate parts of the information in the database using a rule-based
231 approach (Azibi & Vanderpooten 2002). An example of a rule in this context was “if there is
232 no indicator in the database testifying that a given wetland plays a role in flood mitigation,
233 then this wetland is assigned to the category “*No information in the database suggesting that*
234 *it is suitable, even poorly, to pursue this objective to conserve wetlands performing a flood*
235 *regulation function.*” A rule-based approach consists in identifying a consistent set of such
236 rules allowing information in the database to be aggregated. To design these rules, we worked
237 with representatives of wetland manager groups, who collectively identified a series of
238 management objectives that they deemed they had political legitimacy to choose. We then
239 used a rule-based aggregation method and MR-Sort, a non-compensatory aggregation method

240 (Leroy et al. 2011), to produce a framework that the wetland taskforce will be able to use
241 autonomously.

242 (4) The concrete outcome is a platform with which the wetland taskforce will be able to prioritize
243 wetlands for managers, in a legitimate and fine-tuned way, thereby fulfilling the promise to
244 add value and strengthen legitimacy by paying particular attention to the nature and design of
245 data. The increased legitimacy stems from the fact that, whereas standard practices in wetland
246 prioritization indiscriminately conflate technical choices (concerning, for example, the
247 reliability of this or that indicator) and political choices (for example, choices of objectives to
248 pursue), this platform makes a point not to preempt the latter (see Choulak et al. 2019 for
249 more details). The platform has been applied to several projects over the past few months
250 (Melanie Paris, personal communication), and regional-scale funding institutions are
251 interested in applying this new method at a larger scale. From a theoretical point of view, our
252 main contribution is the notion of “meta-decision analysis.” This notion stresses that, while
253 researchers in decision sciences can provide decision support to decision makers in some
254 contexts, many other actors, such as consultants, experts, stakeholders, and so on, can play
255 the role of “decision support providers.” Instead of providing decision support to a particular
256 decision maker facing a particular problem, a researcher involved in “Meta-decision analysis”
257 will strive to identify and help legitimate “decision support providers” to help decision-
258 makers (see Choulak et al. 2019, section 2). Meta decision support is, in our view, a corollary
259 of the emphasis on legitimacy championed by authors in the policy analytics space.

260

261 3.1.2. Case 2: Facilitating dialogue over a marine pollution dispute

262 (1) The second case study relates to the “red mud” conflict in the Calanques National Park (South
263 France). In Marseille, there is an enduring dispute about waste disposal in the Mediterranean
264 Sea, which is supposedly forbidden by the Barcelona convention of 1992. A factory has had a
265 long-term special dispensation allowing it to dispose of massive quantities of residuals of the
266 transformation of bauxite—the so called “red mud”. This pollution is considered illegitimate
267 by a part of the population and creates a strong political conflict, although most people also
268 acknowledge that the jobs provided by this factory are vital for the area. Despite public worries,
269 the administration believes that all has been done to improve practices – but there is no
270 communication among opposing worlds and thus no reduction of political conflict, and as a
271 result the main argumentative discussions take place in judicial courts.

272 (2) In this context, the data available on past and current disputes are numerous (e.g. reports by
273 experts and consultants, surveys by journalists, scientific studies, data from monitoring
274 programs). However, in this deeply conflictual context, some of these data can be easily
275 manipulated, and tracing back the biases that might have plagued them is hazardous. This is
276 why the researcher involved in this case study saw it as an especially potent illustration of the
277 idea, stressed in Tsoukias et al. (2013), that in such a complex context, sui generis processes
278 are required to generate reliable data.

279 (3) A role-playing game was co-produced with local inhabitants, environmental associations,
280 political decision makers and representatives of the factory to represent a range of points of
281 view and values in a single format. Based on long interviews, cognitive maps that brought
282 together definition of problems, actors, and possible actions were produced. Lastly, three
283 participatory techniques were used to help structure debates: a serious game, participatory
284 theatre and the co-construction of a research project between researchers and activists. The

285 serious game initially aimed to create debate but was transformed into an education game
286 because the field study itself created too much tension. It has been used in diverse contexts in
287 the region since then, but never with a group of people in serious conflict. Artists then
288 developed a theater play to organize discussion forums where opponents to the factory,
289 involved scientists, and the general public met and generated new discussions about the
290 problem and the possibilities for solving it. Eventually 50 interested people were invited to co-
291 construct a new research project about the multiplicity of forms of pollution and their
292 circulation in the area, so as to raise awareness of the red mud issue and evaluate the
293 vulnerability of the territory.

294 (4) The outcomes of this case study confirm the fruitfulness of developing *sui generis* tools
295 generating entirely new data, in a context in which analyzing existing data would be
296 methodologically questionable. The continued adaptation of the choice of participatory
297 techniques and their implementation in this case helped to better understand the diversity of
298 points of view. Contradictory normative views concerning social priorities could be
299 characterized and discussed, which facilitated communication among opposing worlds. The
300 co-constructed knowledge production has strengthened links between scientists and
301 associations, who in parallel have found representatives able to interact regularly with the
302 administration. Public trust in the administration was thereby strengthened and the
303 administration renewed their interest in creating arenas of dialog. However, the political
304 problem lingers on.

305

306 3.1.3. Case 3: Facilitating reflection on a collaborative water management network

307 (1) The third case focuses on the construction of collaborative environmental networks in the
308 Gironde estuary (New Aquitaine, South West France) (Boschet & Rambonilaza 2017). In the
309 context of the Water Framework Directive (WFD) and its implementation at the local river
310 basin scale, as well as the Birds and Habitat Directives (Natura 2000 sites), several
311 participatory mechanisms have been introduced. At the same time, local decision-makers have
312 expressed their wishes to orient the future development of the riparian municipalities around
313 the preservation and enhancement of natural and heritage resources, in an area that has
314 historically had vocation as an industrial port. The major challenges were the lack of links
315 between the two shores of the estuary, and a lack of visibility for the group of stakeholders
316 who deal with the environmental issues of the estuary.

317 (2) One of the most important ideas emphasized by Tsoukias et al. (2013) in their introductory
318 definition of policy analytics is that public policy contexts make it particularly difficult to use
319 the sophisticated techniques typically associated with so-called ‘analytics.’ This is due to the
320 fact that these sophisticated techniques are difficult for stakeholders and decision makers to
321 understand, whereas in public policy contexts, transparency, participation and deliberation
322 play a key role. The researchers involved in the present case study saw this context as an
323 opportunity to test if it is possible to meet both policy analytics ambitions, by putting some
324 sophisticated analytic techniques—in this case network analysis and statistical models—to
325 use to help actors understand their interactions and to coproduce new interactions.

326 (3) The case study involved an ex-post analysis of the functioning of collaborative environmental
327 governance and the main factors explaining how collaboration relationships form, and an
328 assessment of the heterogeneity and representativeness of the stakeholders involved, as
329 recommended by WFD (Art. 14). Data collection used documentary sources to identify

330 representatives of organizations and count their participation in four policy processes in the
331 Gironde estuary (514 individuals representing 386 organizations). A two-mode network
332 methodology and preliminary field survey was used to define the population of interest (“the
333 actors who act”). Then a final survey of this population produced data covering their
334 exchanges of information, expertise and resources, as well as the names of the people who are
335 members of their network, who were themselves interviewed afterwards. The interviewees
336 were asked to name the network members who are the most important in the estuary’s
337 environmental management, then in a second step to name their actual partners, leading them
338 to distinguish their understanding of the whole network and their personal network of
339 collaboration. The survey, which followed a snowball sample methodology, was halted when
340 no new names were mentioned by the respondents. These questions were integrated into a
341 broader interview grid, which highlighted the interviewees' perceptions of opportunities and
342 barriers to working with potential partners. The use of data (the actors involved and their
343 relationships) first provided the current state of the collaborative network: the actors and their
344 links, their position in the network, and the diversity of exchanges (financial, informational,
345 contractual, informal...). A second step, which used statistical models of networks, consisted
346 of assessing the factors facilitating or enabling collaboration links. In particular, the distance
347 between the actors was systematically analyzed. By “distance”, we mean not only physical
348 distance, but also institutional distance (the positioning of stakeholders in relation to the rules
349 governing the management of environmental issues); organizational distance (the principles
350 that dictate the involvement of stakeholders within participation devices); and finally statutory
351 distance (the specificity introduced by the roles devolved to the political and administrative

352 apparatus via the statutes of the actors, elected or bureaucrat). The outcome was a
353 visualization of the collaborative network.

354 (4) This work makes several contributions, illustrating how analytics tools can be put to use in a
355 public policy context, despite the prima facie contradiction between the complexity of these
356 tools and the requirements of participation. It provides a robust representation of the current
357 state of the group of actors involved and a factual proof of the separation between the two
358 shores in terms of collaboration, and cognitive support to the actors involved in terms of their
359 social working environment. It also helped the Gironde and Charente local administration
360 (“Conseils Départementaux”), and the “Syndicat Mixte” of the Gironde Estuary, to rebuild
361 the collaborative network of actors mobilized around environmental stakes in the estuary. It
362 is also a renewal of the political economy analysis of the implementation of environmental
363 policies at the local level. This work also forced some actors to acknowledge the inertia of
364 some networks of interaction, and its adverse implications. This eventually enticed them to
365 encourage the arrival of new entrants, particularly economic players who have developed
366 activities related to the estuary's heritage.

367

368 3.1.4. Case 4: Water management policy design

369 (1) This case study deals with water management in the agricultural system of the Apulia Region
370 (Italy), characterized by policy resistance that hampers the implementation of a water
371 protection policy. Due to the limited availability of water resources, the agricultural activities
372 are characterized by the combined use of both surface water and groundwater. Groundwater
373 overexploitation depletes water quantity and quality, leading to long term social and
374 environmental problems, including restrictive groundwater measures according to the Water

375 Framework Directive (Portoghese et al., 2013). The policies implemented in the area aim
376 either to improve the efficiency of groundwater use through innovative irrigation techniques
377 or to restrict groundwater use through policies and a tight control of Farmers' activities
378 (Giordano et al. 2015). Based on a traditional policy making approach, this policy was
379 developed without considering the potential impacts on the stakeholders, creating strong
380 conflict between stakeholders. This case study hence represents an emblematic example of
381 the complexity of water management, where decision-makers with competing objectives and
382 values need to share the same resource. A limited understanding of the different problem
383 framings can be a source of conflict, hampering the implementation of and/or reducing the
384 effectiveness of environmental policies (Giordano et al. 2017). Stakeholders act as if the
385 decision space was as simple as they presume it to be (i.e. ignoring the role of some of the
386 other actors and/or making assumptions about their decisional processes). A detailed
387 description of the case study and the analysis of the ambiguity in problem framing can be
388 found in Giordano et al. (2017) and Pluchinotta et al. 2018.

389 (2) By highlighting the distinctive challenges involved in trying to use 'analytics' tools in public
390 policy contexts, publications on policy analytics provide a partial explanation of the fact that
391 sophisticated decision support methods tend to be poorly used at least in some public policy
392 contexts. It occurred to the authors involved in this case study that their context of defective
393 water management policies illustrated this idea. They therefore took this context as an
394 opportunity to try to fulfil the corresponding promise of policy analytics, which is to put state-
395 of-the-art decision support tools to use in a complex and conflictual public policy context.

396 (3) The data generating work focused on the policy design process (i.e. design of policy
397 alternatives), using an innovative participatory approach. Mainstream policy tends to neglect

398 the generation of novel policy alternatives and is more concerned with evaluating known
399 alternatives (Ferretti et al. 2018, Pluchinotta et al. 2019). The experiences carried out in the
400 Apulia case study supported the application of the Policy-KCP participatory tool for the
401 design of policy alternatives, integrating Decision science and Design theory. Policy-KCP (P-
402 KCP) is a Concept–Knowledge theory driven tool (i.e. one of the available design theories),
403 adapted to the design of abstract objects such as public policies. The P-KCP aims to formalize
404 the innovative design of policy alternatives within a public decision-making process. The P-
405 KCP supports the creation of a shared artefact (Ostanello and Tsoukiàs 1993), further
406 motivating stakeholders’ engagement and commitment to a participative policy making
407 process. The steps of the P-KCP participatory tool are described in Pluchinotta et al. (2019).
408 The P-KCP participatory tool assisted policy makers and stakeholders to work together to the
409 generate policy alternatives and overcome the difficulties of traditional approaches. The phase
410 of knowledge elicitation and alignment (P–K phase) represents the starting point for building
411 a shared concern, toward a generative phase (P–C phase). The P–K phase supported
412 identification of the state of common knowledge on groundwater protection and surface water
413 management problems, including the quali-quantitative state of aquifers and the analysis of
414 the different stakeholders’ problem framing (Giordano et al 2017). The knowledge elicitation
415 activities were carried out by integrating scientific and technical pieces of evidence available
416 in literature with expert and local knowledge according to participatory work principles. The
417 results of semi-structured interviews structured in mental models were combined with the
418 outputs of the stakeholders’ analysis and scientific literature studies, available data, emerging
419 technologies, best practices and current policies.

420 (4) The main outcome of this study was the pilot application of an original approach for the
421 innovative design of policy alternatives, illustrating how a state-of-the-art decision support
422 tool can be implemented in a complex and conflictual public policy setting. The proposed
423 methodology (P-KCP), integrating Decision Science and Design theory, formalized the policy
424 design process and supported the generation of previously unimaginable policy alternatives.
425 It connected local and expert knowledge within the whole design process thanks to the
426 construction of a collective problem understanding (i.e. a shared concern). It brought together
427 stakeholders, experts, institutional and non-institutional actors aiding them to find new ways
428 of working together efficiently, generating innovative possible alternatives and encouraging
429 longer term thinking. As a result, we observed that policy design can be a generative process
430 for the creation of a new dimension of values, through the creation of new variables and/or
431 the elimination of variables having little value for the process. For example, within the case
432 study, we were able to introduce new alternatives in order to modify the value structures in a
433 successful policy making.

434

435 3.1.5. Case 5: Decision support for catchment management

436 (1) This study deals with a collection of decision support processes involving modelling for
437 integrated catchment management and the stakeholders of these catchments, carried out by a
438 team of researchers at The Integrated Catchment Assessment and Management Centre
439 (iCAM) at the Australian National University over the past few decades (see Merritt et al.
440 (2017) for an overview of some applications). Integrated Water Resources Management is a
441 widely recognised paradigm for making more inclusive policy decisions regulating the many,
442 often competing, users of water; however, without effective decision support or ‘policy

443 analytics' the promise of the paradigm is hard to realise. Focusing on a typical situation, a
444 project is developed in partnership with water management authorities in Australia through
445 co-creation of a research topic, informed by both opportunities identified by the university
446 and available resources and priorities of the agency. To ensure legitimacy of the decision-
447 support processes and models, a steering committee is used to provide feedback, in addition
448 to having close involvement from government personnel and landholders.

449 (2) While some of the collection of work in this case study pre-dates discussion of the expression
450 "policy analytics", the researchers involved consider the use of analytical tools to support
451 policy decision making eminently aligned with policy analytics, notably through the use of
452 participatory techniques combined with integrated modelling; the projects typically satisfy all
453 four normative principles defining policy analytics, as listed in the next section.

454 (3) A typical project merges data and information from stakeholders and science through
455 participatory processes and integrated modelling. Modelling provides a natural means for
456 organizing and integrating economic, ecological, hydrological data, qualitative stakeholder
457 input and interviews. An iterative process is adopted (Jakeman et al. 2006), recognising that
458 design of both participatory processes and integrated models needs to be purpose and context-
459 driven, but that new information arises over time that require changes to the project plan
460 (Lahtinen et al. 2017). Data used in the construction of models and that from their resulting
461 outputs play an important role in water management in understanding biophysical processes
462 and anticipating impact of policy or management measures. Integrated modelling then helps
463 to tie economic and ecological outcomes with hydrological processes and intervention
464 measures. Workshops to gain a common understanding of the system are supplemented by
465 interviews targeting sector-specific understanding of agriculture and ecological outcomes. A

466 pragmatic model building approach is used, involving representing systems at the required
467 level of complexity and mixing methods for different model components in order to best
468 integrate knowledge of decision makers, multiple expert disciplines, and on-the-ground
469 stakeholders. A spatially semi-distributed hydrological model provides information at key
470 points and aggregate regions, reducing the risk of information overload for users, and allowing
471 for interactive use of the model. Uncertainty in outcomes is dealt with using scenarios and
472 Bayesian Networks (Kelly et al. 2013; Maier et al. 2016), which have typically received
473 positive feedback from users. The result is inherently interdisciplinary, such that
474 communication within the project plays an important role.

475 (4) Project outcomes are delivered both through the stakeholder engagement process and
476 produced model and decision support tool. The stakeholder engagement process facilitates
477 social learning and shared problem framing, as well as building trust in the model. The model
478 provides cross-sectoral estimates of the impact of various water policies and management
479 interventions, in a transparent, traceable manner that the stakeholders can critique and discuss.
480 Both the engagement process and produced tool then influence regulatory and agricultural
481 decision making processes. Importantly, there is no ex ante expectation that the model or
482 outputs are directly *referenced* in decision making. The project is understood to be one of
483 many sources of information that decision makers draw on. Shaping understanding of the
484 situation is the main priority, along with adjusting different stakeholders' views of how the
485 world operates and their relationships to each other, which makes evaluation of this type of
486 policy analytics project particularly challenging (Hamilton et al. 2019).

487

488 3.1.6. Case 6: participatory revision of a water management plan

489 (1) This study focuses on the participatory process used to revise a water management plan in the
490 Drôme river valley, located in southeastern France. The river is managed by a basin institution
491 and a local water committee. The basin institution is in charge of coordinating stakeholders,
492 facilitating the local water committee and carrying out construction and maintenance work.
493 The local water committee is in charge of developing, revising and monitoring the
494 implementation of the river management plan. The first river management plan in the Drôme
495 was established in the mid 90's (the Drôme was the first river basin in France to establish a
496 river management plan). This plan was revised for the first time between 2007 and 2013. For
497 the second revision, starting in 2018, policy-makers were willing to use an innovative
498 approach, by enabling citizens to make concrete proposals that would then be examined by
499 the local water committee for inclusion in the revised river management plan. This
500 participatory process was supported by a European project, SPARE (Strategic Planning for
501 Alpine River Ecosystems, co-financed by the European Union via Interreg Alpine Space), and
502 by international researchers. As a result, between November 2016 and October 2018, 344
503 citizens were involved in the: i) launching of the process, ii) design of the process, iii)
504 participatory diagnosis of the river basin, iv) identification of main stakes of the river basin
505 and proposing of actions and v) synthesis of the results. In total, 62 participatory events were
506 organized over 2 years.

507 (2) The researchers involved saw this context as an opportunity to explore an aspect of the
508 ambition heralded by policy analytics: how a large amount of data could be gathered and
509 analyzed in a participatory context, in such a way as to improve the decisions made by policy
510 makers by anchoring them in new data, while monitoring the involvement of participants in
511 the process.

512 (3) The various steps of the process produced a large amount of data, including 85 initial
513 questionnaires about citizens' perceptions of the river and of participation, 630 contributions
514 to the citizen diagnosis, 189 propositions of actions, 3 action plans, 1 final report, 5 thematic
515 syntheses sent to the local water committee and answers to 78 questions asked by citizens. In
516 addition, the participatory process itself was monitored and evaluated to provide data about:
517 the composition of the participants' group, its representativeness, the retention level of
518 participants (whether participants stayed throughout the whole process or left part way
519 through), etc. Data were collected by researchers, facilitators and participants themselves. A
520 group of 16 citizen volunteers contributed to data framing and collection. Data analyses were
521 made by researchers and policy makers while the process was underway.

522 (4) The project facilitated a better understanding of the opinions, values and judgements of
523 participants: for example, the 85 initial questionnaires provided data about citizens'
524 perceptions of the river and of participation (see results in Ferrand, Girard, & Hassenforder,
525 2018 and <http://www.alpine-space.eu/projects/spare/en/pilot-case-studies/drome/charts>). The
526 participatory diagnosis outlined what participants liked or disliked in the river basin, what
527 they considered needed to be conserved or modified, what data they lacked and what questions
528 they had. The results were also used to support policy makers, at two levels. First, data
529 produced by participants fueled the revision of the water management plan. It highlighted
530 issues that were important to citizens and that had been to date left out by policy makers, a
531 number of which were subjective, such as the importance of the landscape, and attachment to
532 the territory. It also allowed an analysis of who was present during the process and those who
533 were absent. For instance, since the process attracted mainly people over 65 in its initial
534 phases, an online participatory tool was set up for the action proposal phase so that working

535 people and parents could participate as well. As a result, 52 additional participants
536 contributed. Adapting the process in real time illustrated how data gathering and analysis can
537 be included in the participatory process, rather than being postponed to the end the
538 participatory phase. Following a similar adaptive logic, the analysis of the participants' group
539 composition also fueled the reflexivity of the group of participants who wondered whether
540 they were legitimate to make decisions about the river if they were not representative of the
541 population. Finally, the project strengthened the policy process in the sense that all the data
542 produced was proof-read by participants and then put online, thereby improving the overall
543 transparency of the policy-making process (results were presented during participatory events
544 and available online on a forum that was set up purposefully:
545 <https://sites.google.com/site/dromenjeu/>). As a result, newcomers could see what had been
546 produced by the group when they joined the process, and participants could promote and share
547 their productions.

548

549 3.2. Properties of applications of policy analytics

550 As detailed in section 2, discussions and reflections on the above case studies (and additional ones
551 which are not detailed here, such as Kana et al. 2014, Merritt et al. 2017 and Raboun et al. 2019),
552 led to the collective identification of normative properties that, we claim, should accompany
553 applications of policy analytics. The case studies explored above do not specifically embody all
554 these properties since they were not designed with these properties in mind. Rather, they were
555 motivated by publications and discussions on policy analytics or by ideas that featured prominently
556 in such discussions. The properties in this section were thus identified *ex post* from the collective

557 analysis of these studies. Future works embodying our four normative properties will demonstrate
558 what we now consider to be important attributes for policy analytics approaches. The first two
559 properties are concerned with capturing the specific aspects of policy analytics associated with its
560 anchoring in decision analysis. The other two are meant to outline policy analytics features
561 associated with its application to public policies.

562 We do not claim that each one of these properties is entirely novel for public policy studies. Many
563 studies could rightfully claim that they satisfy one of these properties, and there might even be
564 applications which satisfy several of them. Our claim is that a study that satisfies them all
565 materializes the ambition underlying the policy analytics research program.

566

567 P1: Demand-orientedness. Our experiences in the different case studies above showed us that, in
568 most cases, the fact that our academic initiatives could easily respond to a demand voiced by actors
569 in the field was key to fulfill the ambition of co-producing solutions with decision makers. In the
570 various cases in which the project was directly and explicitly requested by an institution or an actor
571 (the wetland taskforce and, ultimately, the consortium of water related institutions in case 3.1.1,
572 the local regional administration in case 3.1.3, various water management authorities in case 3.1.5,
573 the basin institution in case 3.1.6), this strengthened the involvement of various actors in the
574 decision process, including of course the one issuing the request but with others as well. In the
575 other cases (3.1.2 and 3.1.4), although the project stemmed from an initially academic questioning
576 point of view, the fact that they were addressing problems that actors deemed important played a
577 key role, which was demonstrated by the fact that various actors ultimately endorsed the
578 questioning as their own. This suggests the importance of endorsing the normative idea that the

579 justification of, and motivation for, an application of policy analytics should not be purely
580 academic, and should be anchored in a real demand, voiced by actors, groups or institutions in the
581 field. This does not always mean that the demand should pre-exist and be voiced by an actor or
582 institution already enjoying a form of authority: it can be created as the research project unfolds,
583 which can take time. But in that case the created demand will qualify as a demand properly
584 speaking, and the study will qualify as demand-oriented, if and only if there are actors or groups
585 or institutions who end-up endorsing this demand and making use of the approach and its
586 outcomes. This theoretically disqualifies academic studies that do not respond to an actual use
587 case, even if they claim to respond to a generic “societal demand”. We note that there will be much
588 useful academic work required that may be pre-cursory to being able to apply policy analytics
589 approaches in a demand-orientated manner, such as algorithm development and other
590 methodological developments; and that in such situations the distinctions between good theory
591 development and praxis in any application-focussed academic endeavor are inherently fuzzy.

592

593 P2: Performativity. By promoting operationalisation and the importance of co-production, policy
594 analytics stresses that decision support interventions should not be purely academic, and should
595 rather feed concrete applications, leading to improvements of the situation they study. This idea
596 played a key role in all of our case studies: in case 3.1.1, the outcome was a new prioritization tool
597 that the decision aiding provider will use on a daily basis in its interactions with wetland managers,
598 which will inevitably lead to concrete changes in their conservation strategies and in the concrete
599 restoration actions they will implement. In case 3.1.2, the project deployment led to the
600 construction of an active debate arena, enabling discussions among concerned populations to be
601 reorganized. The analytical results in case 3.1.3 helped to guide future actions of decision-makers

602 in association with the actors of the collaborative network, leading to the emergence of a new
603 "policy trajectory". In case 3.1.4, the study designed new policy alternatives, which will be
604 included in and enrich existing policy making processes. In case 3.1.5, water managers in
605 numerous settings used the results of the modelling exercise to inform and make planning
606 decisions. In case 3.1.6, the intervention led to process adaptations as illustrated by the online
607 participatory tool set up for the action proposal phase. In all cases, this direct link with applications
608 played a key role in ensuring the relevance and operationality of the approach. This suggests the
609 following normative property: the aim of applications of policy analytics should not simply be to
610 describe or analyze states of affairs or processes; it should be to support actions which will
611 encourage improvements of these states of affairs and processes, ideally in new and positive
612 directions. This application-focused aspect is what we call "performativity". This excludes purely
613 descriptive approaches. However, it does not exclude integration of descriptive sub-studies within
614 a policy analytics project.

615

616 P3: Normative transparency. Our various case studies show that, when trying to fulfil particular
617 aspect of the initial policy analytics' ambition, we were all led to work out our own normative
618 assumptions and forced to clarify and display them. This involves amongst others: reflexively
619 identifying or choosing the role that analysts have in their interactions with decision-makers
620 (illustrated in particular in case 3.1.1); analyzing and improving existing decision aiding structures
621 (3.1.3); analyzing and modifying when needed the set of stakeholders, concerned citizens, and
622 various experts that are involved in the decision process (3.1.6); analyzing the broader significance
623 of the results of the study, and its chosen boundaries, to identify if and how they can support more
624 generalized conclusions (3.1.1, 3.1.3, 3.1.4, 3.1.5, 3.1.6). This requirement was present from the

625 start in case 3.1.1, since the data was specifically selected and aggregated in such a way as to
626 prevent any risk that some actors might think that the method used preempted legitimate political
627 or other value-laden choices. In case 3.1.2, normative considerations did not take center stage at
628 the beginning of the project, but because the first results unveiled clashes of normative frameworks
629 among the actors concerned, the need to be transparent with respect to the normative underpinning
630 of the methods used ended-up playing a key role. In cases 3.1.3 to 3.1.6, the participatory aspects
631 of the study similarly led to the emergence of a diversity of value frames, which had to be taken
632 into account on an equal footing, thereby forcing our own interventions to be transparent with
633 respect to their normative anchorage. With the benefit of hindsight, this idea appears crucial, since
634 it conditions our ability to support decision makers in their own attempts to be transparent and
635 accountable, in particular in their interactions with decision support providers (be they researchers,
636 consultants or in-house policy analysts). This suggests the following normative property:
637 applications of policy analytics should clarify, display and account for their normative
638 underpinnings, both in terms of the points of view taken into account and in terms of how
639 interactions between analysts, decision-makers and stakeholders unfold. This property excludes,
640 for example, welfarist economic, public management approaches and others that do not make
641 explicit their ethics and values-based assumptions.

642

643 P4: Data meaningfulness. The term “analytics”, in “policy analytics”, was purposefully chosen to
644 emphasize that one of the most important (if not the most important) ambitions of policy analytics
645 is to reinforce the importance of reflecting on the nature and meaning of data used to support
646 policies. The general availability of numerous and sometimes large datasets that characterizes our
647 digital age means that large quantities of data can be easily accessed and computed. But

648 information on the context that has led to the emergence of these data, the protocols used, their
649 intrinsic limits, the paradigms that should accompany their interpretation; rather than being
650 forgotten in this process. Devictor & Bensaude-Vincent (2016) and Jaric et al. (2019) provide
651 detailed examples of the problems that this can create for environmental policies, as data are
652 computed and interpreted in questionable ways. Several of our case studies were motivated by
653 attempts to master the whole process of data generation and analysis needed to overcome such
654 problems. In case 3.1.1, data were specifically selected and aggregated in different ways depending
655 on how stakeholders understand them. The choice of aggregation methods was then dictated by
656 the interpretation of the data shared among acknowledged experts, and known or suspected
657 associated uncertainties and knowledge-gaps, which involved avoiding commonly used, more
658 mechanistic weighted-sum methods which silence these features of data. In case 3.1.2, the methods
659 used guided the data collection rather than the other way around. In case 3.1.3, the data were
660 constructed with the actors with a continuing attention to how various actors or groups understood
661 them. In case 3.1.4, the Policy-KCP participatory tool (Pluchinotta et al. 2019, Giordano et al.
662 2020) assisted collaboration between policy makers and stakeholders, connecting local and expert
663 knowledge within the whole design process thanks to the construction of a collective problem
664 understanding (i.e. a shared concern). Similarly, in cases 3.1.5 and 3.1.6, participants were
665 encouraged to contribute to data framing and collection (P3). In all the cases, the data
666 meaningfulness issue hence appears crucial, and the *ex post* analysis even suggests that it could
667 have played a more central role. This is why we champion the following normative property: the
668 analysis of the nature and meaning of data, determined by their context of emergence, protocols
669 used, intrinsic uncertainties and limits, and associated paradigm, should all play a key role in any
670 application of policy analytics. Notice that this requirement does not prevent including, and even

671 advocating for, gathering experience on the go, for example through using real-time sensor feeds
672 or logbooks. These tools are meaningful for both reflexive ex post analysis and formative tracking
673 of system impacts, providing some immediate reflexivity or ‘feedback’ to be used in the policy
674 process itself, for example to identify a particular threshold that may be crossed.

675

676 The four properties articulated here can thus be seen to provide a concrete shape to the promise of
677 policy analytics approaches, including to allow them to tackle a number of challenges associated
678 with digital age and participation, as spelled out in the introduction. Data meaningfulness (P4)
679 aims to reduce the risk of policy makers feeling overwhelmed by data, whose analysis can end-up
680 being entirely beyond their control, as well as to allow them to benefit from messy or unstructured
681 data produced through participatory processes. Normative transparency (P3) can similarly be seen
682 as a safeguard to prevent decision processes from being captured by blackbox models and policy
683 processes that obfuscate the actors and their stakes or interests in them. These two properties can
684 be seen as two constraints on decision support activities that, in what might seem to be paradox at
685 first glance, are at the same time all the more important and all the more difficult to abide by in
686 the digital age. The importance and difficulty of the challenge justifies the need for not just
687 incremental improvements in policy analytics practice, but also major, disruptive innovations in
688 policy making. These can only be delivered by ambitious research activities rethinking the very
689 structure of decision support science and practice. This is epitomized again by the emphasis on
690 learning in P3 (normative transparency), while emphasizing that the innovations produced should
691 have impacts in real life (P2, performativity) and fulfill real needs or demands rather than emerging
692 from purely theoretical whims (P1, demand-orientedness).

693 Based on this analysis, we claim that these four normative properties should be understood as a
694 definition for a *bone fide* application of policy analytics. Our case studies were not elaborated with
695 these four normative properties in mind. Rather, as explained in our methodology, they were
696 elaborated with the ambition articulated by policy analytics in mind. Specifically, the properties
697 were ventured *ex post*, through a structured collaboration process of discussion and case study
698 analysis, so as to strengthen applications of policy analytics in the future. The six case studies
699 therefore do not all materialize the four properties to the same degree. The four properties,
700 however, arguably account for important aspects of all six case studies, and point to areas where
701 each could have been ideally improved to lead to greater policy impact.

702

703 4. Agenda for further policy analytics research

704 As the above account illustrates, we conceive of the development of policy analytics as a dynamic
705 project. It was launched as a conceptual contribution, but its contours are being refined as more
706 and more practical applications have been uncovered from past practice, recently implemented
707 with the policy analytics concept in mind, and subsequently stimulated reflection and prompted
708 adjustments to both policy analytics theory and praxis. This article attempted to capture the core
709 ideas and motivations underlying recent applications and developments of the concept. However,
710 the resulting picture should not be seen as a final description, but rather as a step in a continuing
711 dynamic, whereby we hope to further improve the framework in the years to come through new
712 applications to what we see as emergent, challenging and pressing issues. In this final section, we
713 would like to emphasize a handful of the major issues which could structure a useful research
714 agenda for the policy analytics community in the near future to support it to achieve its ambitions.

715 The connection of each research frontier to the properties spelled out above (P1-4) is also briefly
716 discussed.

717 Our examples above highlighted the importance of participatory approaches in demand-
718 orientedness (P1). Accordingly, fully implementing this property raised challenges pertaining to
719 stakeholder selection issues, which have been an important research topic for a long time for
720 researchers concerned with engineering participatory processes and participation in policy
721 decisions (e.g. Daniell, 2012; Nabatchi, 2016). The works developed by policy analytics
722 researchers allowed important advances in the design of participatory processes and continuous
723 diffusion of data and information through these processes so as to ensure transparency, relevance,
724 and informed decision-making. However, as the process unfolds, the boundaries of the issues
725 tackled and problem formulations can evolve. Due to this evolution, the group of stakeholders
726 initially selected can become incomplete or partly irrelevant at a given stage of a policy-support
727 process. Similarly, a choice made initially concerning the process design, e.g. the participatory
728 methods selected or the roles assigned to some participants, may no longer be relevant later given
729 this evolution. There is therefore a need to identify technologies or procedures to (1) facilitate co-
730 evolution of the participants involved and of the process design, while (2) keeping a memory of
731 previous dialogues, achievements and evolutions. This is a major research frontier for which policy
732 analytics' distinctive interest in data analysis and meaning-giving provide value through collection
733 and use of data generated throughout these participatory processes.

734 We have also seen above that participatory aspects of policy analytics projects play an important
735 role in fulfilling the requirements associated with data meaningfulness (P4). Accordingly, another
736 research frontier for the design of participatory processes is to elaborate means of identifying the
737 data and information that the various participants need to meaningfully participate in the decision.

738 Thinking more fundamentally about the notion of data, how data are created, modified, circulated
739 and re-used out of initially designed contexts is also an important challenge, echoing the
740 importance that policy analytics grants to data meaningfulness (P4). This reflection also has
741 aspects concerning data sovereignty and ownership, and what this means for policy analytics under
742 different jurisdictions. Particularly, policy analytics could integrate reflections about issues of
743 power linked to ownership and diffusion of data, or lack thereof. There are also links to issues of
744 data privacy and accessing environmental-related data about people, and how the use of this should
745 be managed. Likewise the challenges of what streams of data can be meaningfully and ethically
746 integrated to provide full (but perhaps too full) a picture of people, their values, interests and
747 preferences is highly topical as governments and corporations look at their data assets and their
748 perceived underuse (e.g. Löfgren and Webster, 2020). More generally speaking on the area of
749 participation linked to policy analytics, and already reported in the literature (Mazri et al. 2019,
750 Daniell et al., 2010), the design of participation structures is itself a topic of participation, requiring
751 design methodologies where participation is pragmatically considered. Data, when used within
752 complex and long decision processes, are generally subject to several manipulation processes.
753 Assuring the quality and meaningfulness of the entire data pipeline is today a major challenge for
754 the whole area of data science (Christophides et al. 2019). An additional critical issue concerning
755 the policy analytics topic is how to introduce innovation within public policies, for example to
756 conceive of currently inconceivable policies. The most promising ideas come from joining
757 analytics with formal design tools, allowing the emergence of “out of the box” designs (Howlett,
758 2011; Pluchinotta et al. 2019), and in some cases a healthy dose of considering science fiction and
759 the cutting edge of artistic inspiration as an options set worthy of formal investigation (Johnson,
760 2011; Wenger et al., 2020).

761 Important research frontiers also concern how to implement normative transparency (P3) in a
762 formalized, rigorous fashion. In this area, formal argumentation theory in artificial intelligence
763 (Rahwan & Simari 2009) holds important promise to help improve discussions around policy
764 analytics interventions. However, the possibility to use these approaches in this setting raises
765 important epistemological and methodological questions that they do not yet tackle. In particular,
766 if these approaches are used in real-life collective decision processes, they will have to answer
767 questions such as: who has the legitimacy to decide which arguments should be seen as good
768 arguments, and which ones should be considered spurious, and how transparency can be
769 guaranteed in argumentation processes? Cailloux & Meinard (2019) proposed a preliminary
770 formulation of a framework designed to overcome this (and other) limitation of such approaches.
771 Important challenges also lie in a proper integration of such tools in the proceedings of discussions
772 among people or groups, and the reflection of individuals involved, which remain the core of what
773 normative transparency refers to.

774 An associated issue, having to do with next generation algorithms (e.g. AI), is related to what
775 metrics are considered relevant when used as part of policy analytics. For example, perhaps
776 explicability of analytical processes and models is less relevant than legibility (Scott, 1998) and
777 trust. This is particularly important in automated/autonomous systems where decision and policy
778 makers may need to understand the different algorithms, data streams and sensors, and hence trust
779 each layer in the supply chain. What would useful policy analytics look like in such systems?

780 Lastly, a major concern for future research that has to do with performativity (P2), is the long term
781 sustainability of the policy analytics interventions. Policy analytics activities should arguably have
782 long term benefits and co-benefits. Hence a future research avenue is to identify what makes policy

783 analytics approaches more salient for long-term policy support and interventions in a variety of
784 contexts.

785 Our six case study examples illustrate how the notion of policy analytics, in its original
786 conceptualization, proved useful to explore important environmental issues and support
787 environmental decision-makers for important decisions in the field. However, this agenda for
788 future research in turn shows how developing the concept in a bottom-up approach, far from
789 closing debates with a final definition, can help to structure future studies and open new research
790 avenues to further strengthen environmental decision support and the application of policy
791 analytics approaches more broadly.

792

793 Acknowledgements

794 This collaborative research was supported by a grant from the ANU Global Research Partnerships
795 Scheme, and two EU Erasmus+ Jean Monnet projects, the ‘Europa Policy Labs’ and the ‘Water
796 Policy Innovation Hub’.

797

798 Literature cited

799 Agogué, M., Kazakçi, A.O., 2014. 10 Years of C–K theory: a survey on the academic and industrial
800 impacts of a design theory, in: Chakrabarti, A., Blessing, L., (Eds), An Anthology of Theories and
801 Models of Design. Springer, London, p. 219-235. https://doi.org/10.1007/978-1-4471-6338-1_11

802 Arts, I, Buijs, AE, Verschoor, G. 2017. Regimes of justification: competing arguments and the
803 construction of legitimacy in Dutch nature conservation practices. *Journal of Environmental*
804 *Planning and Management* 61(5-6): 1070-1084.

805 Azibi, R., Vanderpooten, D., 2002. Construction of rule-based assignment models. *Eur. J. Oper.*
806 *Res.* 138 (2), 274–293.

807 Boschet, C., Rambonilaza, T., 2017. Collaborative environmental governance and transaction
808 costs in partnerships: evidence from a social network approach to water management in France. *J.*
809 *Environ. Plann. Man.* 61, 105-123. <http://dx.doi.org/10.1080/09640568.2017.1290589>

810 Boyd, A., Geerling, T., Gregory, W.J., Kagan, C., Midgley, G., Murray, P. and Walsh, M.P.,
811 2007. Systemic evaluation: a participative, multi-method approach. *J. Oper. Res. Soc.* 58, 1306-
812 1320.

813 Cailloux, O., Meinard, Y., 2019. A formal framework for deliberated judgment. *Theor. Decis.*
814 <https://doi.org/10.1007/s11238-019-09722-7>

815 Choulak, M., Marage, D., Gisbert, M., Paris, M., Meinard, Y., 2019. A meta-decision-analysis
816 approach to structure operational and legitimate environmental policies - With an application to
817 wetland prioritization. *Sci. Total. Environ.* 655, 384-394. doi.org/10.1016/j.scitotenv.2018.11.202

818 Christophides, V., [Efthymiou, V.](#), [Palpanas, T.](#), [Papadakis, G.](#) [Stefanidis, K.](#), 2019. End-to-End
819 Entity Resolution for Big Data: A Survey. [CoRR abs/1905.06397](#)

820 Daniell, K.A., 2012. Co-engineering and participatory water management: organisational
821 challenges for water governance. Cambridge University Press, Cambridge UK.

822 Daniell, K.A, Mazri, C., Tsoukiàs, A., 2010. Real world decision-aiding: a case of participatory
823 water management, in: French, S., Rios-Insua, D., (Eds.), e-Democracy: a group decision and
824 negotiation perspective. Springer-Verlag, Berlin, 125-150.

825 Daniell, KA., Morton, A. Ríos Insua, D., 2015. Policy analysis and policy analytics. *Ann. Oper.*
826 *Res.* 236, 1–13. doi:10.1007/s10479-015-1902-9

827 De Marchi, G., Lucertini, G., Tsoukiàs, A., 2014. From evidence-based policy making to policy
828 analytics. *Ann. Oper. Res.* 236, 15-38.

829 Department of Industry, Innovation and Science (DIIS), 2018. Australia's tech future. Department
830 of Industry, Innovation and Science. [https://www.industry.gov.au/sites/default/files/2018-](https://www.industry.gov.au/sites/default/files/2018-12/australias-tech-future.pdf)
831 [12/australias-tech-future.pdf](https://www.industry.gov.au/sites/default/files/2018-12/australias-tech-future.pdf) (accessed 8 March 2019)

832 Devictor, V., Bensaude-Vincent, B., 2016. From ecological records to big data: the invention of
833 global biodiversity. *HLPS* 38, 13.

834 Federal Data Strategy, 2019. The Federal Data Strategy. <https://strategy.data.gov/> [Accessed 8
835 Mar. 2019] (accessed 8 March 2019).

836 Federal Ministry for Economic Affairs and Energy (FMEAE), 2018. Digital Strategy 2025. Federal
837 Ministry for Economic Affairs and Energy.
838 [https://www.de.digital/DIGITAL/Redaktion/EN/Publikation/digital-strategy-](https://www.de.digital/DIGITAL/Redaktion/EN/Publikation/digital-strategy-2025.pdf?__blob=publicationFile&v=9)
839 [2025.pdf?__blob=publicationFile&v=9](https://www.de.digital/DIGITAL/Redaktion/EN/Publikation/digital-strategy-2025.pdf?__blob=publicationFile&v=9) (accessed 6 March 2019)

840 Ferretti, V., Pluchinotta, I., Tsoukiàs, A., 2018 Supporting decisions in public policy making
841 processes: generation of alternatives and innovation. *Eur. J. Oper. Res.* 273, 353-363.
842 <https://doi.org/10.1016/j.ejor.2018.07.054>.

843 Giordano, R., D'Agostino, D., Apollonio, C., Scardigno, A., Pagano, A., Portoghese, I.,
844 Lamaddalena, N., Piccinni, A.F., Vurro, M., 2015. Evaluating acceptability of groundwater
845 protection measures under different agricultural policies. *Agr. Water. Manag.* 147, 54–66.
846 <https://doi.org/10.1016/j.agwat.2014.07.023>

847 Giordano, R., Brugnach, M., Pluchinotta, I., 2017. Ambiguity in problem framing as a barrier to
848 collective actions: some hints from groundwater protection policy in the Apulia Region Group.
849 *Decis. Negot.* 26, 911-932 DOI 10.1007/s10726-016-9519-1

850 Giordano, R., Pluchinotta, I., Zikos, D., Krueger, T., Tsoukiàs, A., 2020. How to use ambiguity in
851 problem framing for enabling divergent thinking: integrating Problem Structuring Methods and
852 Concept-Knowledge theory, in: White, L., Kunc, M., Malpass, J., Burger, K. (Eds), *Behavioral*
853 *Operational Research: A Capabilities Approach*. Palgrave Macmillan publishers, Basingstoke
854 (UK), pp. 93-117

855 Habermas, J. 1985. *The Theory of Communicative Action*. Beacon Press.

856 Habermas, J. 1990. *The Philosophical Discourse of Modernity*. MIT Press.

857 Hamilton, S.H., Fu, B., Guillaume, J.H.A., Badham, J., Elsawah, S., Gober, P., Hunt, R.J.,
858 Iwanaga, T., Jakeman, A.J., Ames, D.P., Curtis, A., Hill, M.C., Pierce, S.A., Zare, F., 2019. A
859 framework for characterising and evaluating the effectiveness of environmental modelling.
860 *Environ. Model. Softw.* 118, 89-98. <https://doi.org/10.1016/j.envsoft.2019.04.008>

861 Hatchuel, A, Weil, B., 2003. A new approach of innovative design: an introduction to CK theory.
862 In: XIVth international conference on engineering design, 19th–21st August 2003, Stockholm.

863 Hatchuel, A., Le Masson, P., Weil, B., 2009 Design Theory and Collective Creativity: a
864 Theoretical Framework to Evaluate KCP Process. In: International Conference on Engineering
865 Design, ICED'09, 24-27 August 2009, Stanford CA.

866 Howlett, M., 2011. Designing Public Policies: Principles and Instruments. Routledge, London.

867 Jaric, I., Quétier, F., Meinard, Y., 2019. Procrustean beds and empty boxes: on the magic of
868 creating environmental data. *Biol. Conserv.* 237, 248-252.

869 Jakeman, A.J., Letcher, R.A., Norton, J.P., 2006. Ten iterative steps in development and evaluation
870 of environmental models. *Environ. Model. Softw.* 21, 602–614.
871 <https://doi.org/10.1016/j.envsoft.2006.01.004>

872 Jeanmougin, M., Dehais, C., Meinard, Y., 2017. Mismatch between habitat science and habitat
873 directive: Lessons from the French (counter-)example. *Conserv. Lett.* 10, 634-644.

874 Johnson, B.D., 2011. Science Fiction Prototyping: Designing the Future with Science Fiction,
875 Morgan & Claypool Publishers, San Francisco.

876 Johnston, E.W. (Ed), 2015. Governance in the information era: Theory and practice of policy
877 informatics. Routledge, New York.

878 Kana, V., Somé, B., Tsoukiàs, A., 2014. A new methodology for multidimensional poverty
879 measurement based on the capability approach. *Socio-Economic Planning Sciences* 48: 273-289.
880 DOI: 10.1016/j.seps.2014.04.002.

881 Kelly, R.A., Jakeman, A.J., Barreteau, O., Borsuk, M.E., ElSawah, S., Hamilton, S.H., Henriksen,
882 H.J., Kuikka, S., Maier, H.R., Rizzoli, A.E., van Delden, H., Voinov, A.A., 2013. Selecting among
883 five common modelling approaches for integrated environmental assessment and management.
884 *Environ. Model. Softw.* 47, 159-181.

885 Lahtinen, T.J., Guillaume, J.H.A., Hämäläinen, R.P., 2017. Why pay attention to paths in the
886 practice of environmental modelling? *Environ. Model. Softw.* 92, 74–81.
887 <https://doi.org/10.1016/j.envsoft.2017.02.019>

888 Lazer, D., Pentland, A., Adamic, L., Aral, S., Barabási, A.L., Brewer, D., Christakis, N.,
889 Contractor, N., Fowler, J., Gutmann, M., Jebara, T. 2009. Computational social science. *Science*
890 323(5915): 721-723.

891 Le Masson, P., Weil, B., Hatchuel, A., 2017. *Design Theory - Methods and Organization for*
892 *Innovation*. Springer International Publishing.

893 Leroy, A., Mousseau, V., PirLOT, M., 2011. Learning the parameters of a multiple criteria sorting
894 method. In: Brafman, R., Roberts, F., Tsoukias, A. (Eds.), *Algorithmic Decision Theory*. Lecture
895 Notes in Artificial Intelligence. 6992. Springer, pp. 219–233.

896 Löfgren, K. and Webster, C.W.R., 2020. The value of Big Data in government: The case of ‘smart
897 cities’. *Big Data & Society*, 7(1), <https://doi.org/10.1177/2053951720912775>

898 Long, C., 2019. An uncomfortable time to be in politics (or anywhere with a ‘climate’) December
899 12, *The New Matilda*, [https://newmatilda.com/2019/12/12/an-uncomfortable-time-to-be-in-](https://newmatilda.com/2019/12/12/an-uncomfortable-time-to-be-in-politics-or-anywhere-with-a-climate/)
900 [politics-or-anywhere-with-a-climate/](https://newmatilda.com/2019/12/12/an-uncomfortable-time-to-be-in-politics-or-anywhere-with-a-climate/)

901 Maier, H.R., Guillaume, J.H.A., van Delden, H., Riddell, G.A., Haasnoot, M., Kwakkel, J.H. 2016.
902 An Uncertain Future, Deep Uncertainty, Scenarios, Robustness and Adaptation: How Do They Fit
903 Together? *Environ. Model. Softw.* 81, 154–64. doi:[10.1016/j.envsoft.2016.03.014](https://doi.org/10.1016/j.envsoft.2016.03.014)

904 Mazri, C., Daniell, K.A, Tsoukiàs, A., 2019. Decision Support in Participative Contexts: the
905 Organisational Design Dimension. *International Journal of Decision Support Systems Technology*
906 11, 47 – 80.

907 Meinard, Y., 2017. What is a legitimate conservation policy? *Biol. Conserv.* 2013, 115-123.

908 Meinard, Y., Tsoukias, A. 2019. On the rationality of decision aiding processes. *European Journal*
909 *of Operational Research* 273(3): 1074-1084. doi.org/10.1016/j.ejor.2018.0

910 Merritt, W.S., Fu, B., Ticehurst, J.L., El Sawah, S., Vigiak, O., Roberts, A.M., Dyer, F., Pollino,
911 C.A., Guillaume, J.H.A., Croke, B.F.W., Jakeman, A.J., 2017. Realizing Modelling Outcomes: A
912 Synthesis of Success Factors and Their Use in a Retrospective Analysis of 15 Australian Water
913 Resource Projects. *Environ. Model. Softw.* 94, 63–72. doi:10.1016/j.envsoft.2017.03.021

914 Mergel, I., Rethemeyer, R.K. Isett, K., 2016. Big data in public affairs. *Public Admin. Rev.* 76(6):
915 928-937.

916 Midgley, G., 2006. Systems thinking for evaluation, in: Williams, B., Imam, I. (Eds), *Systems*
917 *concepts in evaluation: An expert anthology*, Edge Press, pp.11-34.

918 Nabatchi, T., 2012. Putting the “public” back in public values research: Designing participation to
919 identify and respond to values. *Public Administration Review*, 72(5), pp.699-708.

920 O'Donnell, E.L., Horne, A.C., Godden, L., Head, B., 2019. Cry me a river: building trust and
921 maintaining legitimacy in environmental flows. *Australasian Journal of Water Resources* 23, 1-
922 13. <https://doi.org/10.1080/13241583.2019.1586058>

923 Ostanello, A., Tsoukiàs, A., 1993. An explicative model of 'public' interorganizational
924 interactions. *Eur. J. Oper. Res.* 70:67–82.

925 Patton, M.Q. 2008. *Utilization-focused evaluation*. Sage publications, London.

926 Peirce, C.S. 1966. *Selected Writings*. Dover Publications.

927 Pluchinotta, I., Pagano, A., Giordano, R., Tsoukiàs, A., 2018. A system dynamics model for
928 supporting decision makers in irrigation water management. *J. Environ. Manag.* 223, 815–824.
929 <https://doi.org/10.1016/j.jenvman.2018.06.083>

930 Pluchinotta, I., Kazakçi, A.O., Giordano, R., Tsoukiàs, A., 2019. Design Theory for Generating
931 Alternatives in Public Policy Making. *Group Decis. Negot.* 28, 341–375.

932 Raboun, O. Chojnacki, E., Duffa, C., Rios-Insua, D., Tsoukiàs, A. 2019. Spatial risk assessment
933 in case of multiple nuclear release scenarios. *Socio-Economic Planning Sciences*, vol. 72, 2019.
934 DOI: 10.1016/j.seps.2019.06.006

935 Rahwan, I., Simari, G.R. (Eds.), 2009. *Argumentation in Artificial Intelligence*. Springer, 2009.

936 Richard, A, Mayag, B, Meinard, Y, Talbot, F, Tsoukiàs, A. 2018 How AI could help physicians
937 during their medical consultations: An analysis of physicians' decision process to develop efficient
938 decision support systems for medical consultations. In: *PFIA 2018*, Nancy, France.

939 Scott, J.C., 1998. *Seeing like a State*. Yale University Press, New Haven.

940 The White House, 2019. Executive Order on Maintaining American Leadership in Artificial
941 Intelligence. The White House. [https://www.whitehouse.gov/presidential-actions/executive-](https://www.whitehouse.gov/presidential-actions/executive-order-maintaining-american-leadership-artificial-intelligence/)
942 [order-maintaining-american-leadership-artificial-intelligence/](https://www.whitehouse.gov/presidential-actions/executive-order-maintaining-american-leadership-artificial-intelligence/) (accessed 8 March 2019).

943 Touret, R., Meinard, Y., Petit, J.-C., Tsoukias, A. 2019. Cartographie descriptive du système
944 national français du financement de la recherche sur projet en vue de son évaluation. *Innovations*
945 59: 205-241.

946 Tsoukias, A., Montibeller, G., Lucertini, G., Belton, V., 2013. Policy analytics: an agenda for
947 research and practice. *EURO Journal on Decision Processes* 1, 115-134.

948 Villani, C., 2018. For a Meaningful Artificial Intelligence: Towards a French and European
949 Strategy. Government of France.
950 https://www.aiforhumanity.fr/pdfs/MissionVillani_Report_ENG-VF.pdf (accessed 6 March
951 2019).

952 Webster, G., Creemers, R., Triolo, P. and Kania, E., 2019. Full Translation: China's 'New
953 Generation Artificial Intelligence Development Plan' (2017). New America.
954 [https://www.newamerica.org/cybersecurity-initiative/digichina/blog/full-translation-chinas-new-](https://www.newamerica.org/cybersecurity-initiative/digichina/blog/full-translation-chinas-new-generation-artificial-intelligence-development-plan-2017/)
955 [generation-artificial-intelligence-development-plan-2017/](https://www.newamerica.org/cybersecurity-initiative/digichina/blog/full-translation-chinas-new-generation-artificial-intelligence-development-plan-2017/) (accessed 9 March 2019).

956 Wenger, A., Dunn Cavelty, M. and Jasper, U., 2020. The Politics and Science of the Future:
957 Assembling Future Knowledge and Integrating It into Public Policy and Governance. In: Wenger,
958 A., Jasper, U., Dunn Cavelty, M. (eds.) *The Politics and Science of Prevision: Governing and*
959 *Probing the Future* (pp. 229-251). Routledge.