

CAHIER DU **LAMSADE**

335

Mars 2013

Policy Analytics:
An Agenda for Research and Practice

A. Tsoukias, G. Montibeller, G. Lucertini, V. Belton

POLICY ANALYTICS: AN AGENDA FOR RESEARCH AND PRACTICE

Alexis Tsoukias¹, Gilberto Montibeller², Giulia Lucertini¹, Valerie Belton³

¹ LAMSADE-CNRS, Université Paris Dauphine, FR

² Department of Management, London School of Economics, UK

³ Department of Management Science, University of Strathclyde, UK

Abstract: In recent years the field of decision analysis has been heavily influenced by the “analytics” perspective, which integrates advanced data-mining and learning methods, often associated with increasing access to “Big-Data”, with decision support systems. This rapidly growing and very successful field of Analytics has been strongly business-oriented since its origin and is typically focussed on data-driven decision processes. In public decisions, however, issues such as individual and social values, culture and public engagement play a much bigger role and, to a large extent, characterise the policy cycle of design, testing, implementation, evaluation and review. From this perspective public policy making seems to be a much more socially complex process than has hitherto been considered by most analytics methods and applications. In this paper we thus suggest a framework for the use of analytics in supporting the policy cycle – and conceptualise it as “Business Analytics”.

Key-Words: analytics, policy analysis, decision analysis, policy cycle, decision support.

1 Introduction

Policy making has been a traditional domain of research and practice, where decision analysts¹ have introduced formal methods aimed at helping policy makers improve their decisions (for a recent survey see De Marchi et al. 2012). In recent years the field of decision analysis has been heavily influenced by the “analytics” perspective, which integrates advanced data-mining and learning methods, often associated with increasing access to “Big-Data”, with decision support systems. This rapidly growing and seemingly very successful field of Analytics has been strongly business-oriented since its origin (see Davenport et al. 2010) and is typically focussed on data-driven decision processes. However, in public decisions issues such as individual and social values, culture and public engagement play a much bigger role (White and Bourne 2007) and, to a large extent, characterise the policy cycle of design, testing, implementation, evaluation and review. From this perspective public policy making seems to be a much more socially complex process than has hitherto been considered by most analytics methods and applications (Almquist et al. 2012; Juntti et al. 2009). This is not to deny the potential for analytics to contribute throughout the policy cycle, including through the growing area of text analytics to analyse social media, as well as the increasing public access to data and tools for analysis, e.g. Google Public Data Explorer², Guardian Data blog³). The intention of this paper is to propose a framework for Policy Analytics within which approaches such as these can be appropriately harnessed.

¹ We use the term decision analyst in the broad sense of any operational research/management science professional.

² www.google.co.uk/publicdata/directory

³ <http://www.guardian.co.uk/news/datablog>

Analysing and supporting the design, implementation and assessment of public policies is not really a new domain. Political scientists have worked in this field for decades (Moran et al. 2006). Equally, economists have focussed much of their research on rational theories of public decision making and formal methods for the ex-ante and ex-post evaluation of public policies (see for instance Dollery and Worthington 1996). Cost-Benefit Analysis (see Dasgupta and Pearce 1972; Nas 1996) is widely used and, perhaps, the best known method for evaluating public policies among both practitioners and researchers. However, Cost-Benefit Analysis is not without contemporary critics (for example, Ackerman and Heinzerling 2004; Adler and Posner 2006) and many other approaches have been developed including, in recent years, Real Options Analysis (see Smit and Trigeorgis 2004; Trigeorgis 1990). Decision Analysis and Operational Research have also developed methods which aim at addressing public policy making (see for instance Pollock et al. 1994). At the same time “analytics” has developed as increasingly an independent domain, both for practice and research, growing out of and bringing together more traditional fields such as statistics, data analysis, data mining, knowledge extraction and machine learning (Davenport and Harris 2007; Davenport et al. 2010; Liberatore and Luo 2010).

An important question arising from the above observations is whether the field of Decision Analysis is giving appropriate consideration to the specific requirements for Analytics within the wide area of supporting Public Policy Making. Thus, the main motivation of this paper is to raise a number of questions and to prompt discussion about the skills and awareness that Decision Analysts need to operate effectively in this domain. In particular:

- Are practitioners involved in supporting decision processes in the public domain appropriately using analytics and, if so, what type of analytics could fit the type of support they use?
- Is decision analysis training appropriate for helping the integration of traditional decision analytic tools and analytics in the area of public policy making?
- Are researchers considering appropriately the relationships between the specificities of public policy making and analytics?

Our paper does not seek to provide an exhaustive reply to the above questions, but will try to provide our perspective on them. We also consider it important to raise and discuss the following assertions:

- Policy Making is a type of decision process with specific characteristics, thus demanding dedicated analytical methodologies.
- Business Analytics does not always fit the requirements most policy cycles demand, particularly regarding the creation of knowledge to support such cycles.
- In order to improve both policy making processes and decision aiding processes we need to integrate data-driven decision making with value-driven decision making.
- Currently, most analytics methods are based on benchmarking and descriptive approaches, while supporting the activities occurring within a policy cycle needs to focus on constructive approaches, including constructive benchmarking and learning.

The paper has the following structure. In section 2, we discuss some key characteristics of policy making processes, distinguishing them from other decision processes. Section 3 reviews the types of policy analyses available to decision analysts and policy makers. We then define what we mean by Policy Analytics in section 4 and suggest opportunities that it can offer to decision analysts. The paper concludes with some suggestions on how this emerging field can be further developed.

2 Key Characteristics of Policy Making

What makes a policymaking process different from any other decision process? Why is policy making not just common decision making (but perhaps a little more complex)? We need to understand what characterises policy decisions, making them unique. First of all, we must recognise that speaking about policy-making means speaking about a “policy cycle” (Lasswell 1956). The policy cycle consists of a set of sequential actions linked by some main “goal” or, more generally, by some common public issue. This cycle is composed by eight major steps: issue identification, defining policy objectives, policy design, policy testing, policy finalisation, policy implementation, policy monitoring & evaluation, policy readjustment & innovation.

Not only are policy makers engaged(or should be engaged) in this policy analysis cycle, they are also confronted by five major complexities inherent to public decision making, which we describe next (see also Dunn 2012; Parsons 1995; Hill 1997; Kraft and Furlong 2007; Davies 2004) .

Use of public resources. Although the use of public resources may appear obvious in this context, it also has many implications. First, many resources, tangibles or intangibles, used within the policy cycle are provided by the government or other public institutions and, under such a perspective, are considered public goods. Second, during a policy cycle, policy makers use resources provided also by those who are not involved in the cycle. Third, public policies allocate (or redistribute) resources and among the beneficiaries we can find people who are not implied in the policy cycle and are potentially considered not to be concerned with it. Fourth, decisions (and thus policies) are “irreversible allocations of resources” opening the way to ethical, moral, intergenerational, social justice, environmental issues of very basic nature (almost ideological).

Multiple stakeholders. Policy cycles are “de facto” participative, as many different actors (being citizens, groups or organisations) feel to be naturally involved, as soon as they discover that one of their private concerns can be associated with that policy cycle. Of course participation can be structured or not, allowed or not, visible or not, formal or informal – but it happens independently from the willingness of who “promotes” that given policy cycle. The result is that within the cycle we need to take into account the different concerns carried by the multiple stakeholders and the resources they carry with them, but we also need to handle the different “languages” practiced by all these stakeholders, the distinctive perceptions they might have of the policy cycle, and the different expectations they present. Last, but not least, we need to consider the potential confusion that occasionally may occur, when a policy cycle does not have a clear “decision maker” due to institutional rules, inconsistent legal frameworks and asymmetries in decision power distribution.

Long-time horizon. A policy cycle usually takes a considerable amount of time in order to be developed and this holds even for non-strategic policies. Moreover, the effects and consequences of a policy cycle may become visible only a long time after the cycle occurred and can hold for even longer periods. This may conflict with the agendas of different types of stakeholders. Policy makers usually have short-term agendas due to the timing of politics. Experts and/or analysts may have medium to long-term agendas due to the specific knowledge they have about the policy issue. Citizens may have agendas varying from very short term to very long ones, depending on the concerns they carry in the policy cycle. Such conflicting agendas and different time frames add further structural uncertainty to the policy cycle besides that generated by the difficulty to predict how social and economic scenarios may evolve in the future.

Legitimation and Accountability. What do policy makers look for while engaged in a policy cycle? They look for legitimation: legitimation for themselves, for their actions within that policy cycle specifically, for the outcomes of the policy making process and for the policy making process itself. Legitimation can be obtained from different sources (the law, tradition, moral standards, knowledge, practice etc.) and is the cornerstone of the rationality developed by each stakeholder involved in the

policy cycle (see Habermas 1990). Under such a perspective, a key aspect in gaining legitimization by the policy making process itself are requirements of increasing the level of participation in the policy cycle, of providing transparency for the decision process (explanations and justifications of the outcomes, clear argumentation etc.), and of demanding more accountability (in the sense of “providing an account of”) from policy makers.

Deliberation. Policy cycles occur in the public domain and, at least part of them, are “public decision processes”. In order to be “public”, decision processes need to establish “deliberations”: those moments of the process where “decisions” are formally adopted, become officially known in the public domain, can be enforced by law, and the allocation of resources linked with the decision becomes irreversible. Deliberations are a crucial part of the policy cycle because they structure the timeline where the cycle occurs. However, this is not linearly perceived by the stakeholders: it becomes more “dense” when deliberations are expected to occur (and immediately afterwards).

Why should decision analysts pay special attention to the characteristics mentioned above? It is easy to note that such characteristics are strictly inter-related. They contribute in characterising how the policy making process is structured and, potentially, allow us to understand how it could be conducted. Under such a perspective, decision analysts are expected to provide some decision support exactly in these cases where one or more stakeholders need to establish their position and actions in the policy cycle, an issue captured by the concept called “Action-Arena” (Ostrom 1986) or “Interaction Space” (Ostanello and Tsoukiàs 1993).

In other terms, decision analysis is expected to be used along the whole policy cycle, from the agenda setting to the assessment of alternative actions and their consequences, up to ex-post evaluation of the whole policy. Under such a perspective, we need both a richer toolkit of methods, which support decision makers along the process, as well as a comprehensive methodology allowing for a coherent structuring of the decision aiding process (see Bouyssou et al. 2000; Belton and Stewart 2002; Tsoukiàs 2007; Tsoukiàs 2008).

3 Analysis for Policy Makers

In this section we suggest that there is a growing demand for policy analysis and briefly review different types of modelling already developed with this intent. We then review, again briefly, the main components of the analytics movement.

3.1 Demand for Policy Analysis

Policy decisions impact large numbers of citizens on many different aspects of our social, economic and cultural life. As for any type of decision process, policy makers have to assume the full responsibility of their policies with respect to both private and public stakeholders. Under such a perspective, as we argued before, policy-making is a complex decision process and has always been a field where decision support has been sought: from using statistical information to applying decision analytic and operational research methods (Larson and Odoni 1981; Rosenhead 1981; Pollock et al 1994; Dorling and Simpson 1999).

However, in recent years we can observe a number of trends affecting the nature of demands policy makers address to analysts of any type (for a detailed discussion see Hill 1997; Kraft and Furlong 2007; Moran et al. 2006; Nutley et al. 2003):

- an increasing demand for participation in the policy making process, coming from opinion groups and single citizens;
- an increasing mistrust between citizens and policy makers, as well as between citizens and “experts”;

- an increasing social fragmentation, resulting in a loss of representativeness by traditional political parties and social organisations (such as the trade unions);
- an increasing mistrust of science and consequent limits on its ability to convince citizens about policy consequences and impacts;
- an exponential increase to the amount of information to which citizens can have access, information provided (most of the times) without any check on reliability and truthfulness..

The result is that the policy making process became ever more challenging, since the demand for accountability and legitimation, for both the process and its outcomes, becomes stronger.

To some extent this situation is captured by the appearance in recent years of several manuals and guidelines concerning the assessment of public policies at national and European level, such as:

- the Cost-Benefit Analysis manual of the European Union⁴;
- the Evalsed manuals concerning the assessment and use of the European Social Fund⁵;
- the Green and Magenta Books of the UK Government⁶;
- the Public Policy Assessment Book of the UK Government⁷;
- the Italian law concerning the environmental assessment of Public Works⁸;
- the French law concerning the Technological Risk Assessment Plans⁹.

These documents extend a well-established tradition in the USA related to the use of cost-benefit analysis on policy decisions (for an account of using CBA in the United States regulatory process see Shapiro and Morrall 2012; see also the World Bank manuals about CBA¹⁰).

3.2 Modelling for Policy Making

The use of formal analysis in supporting policy making has a long tradition in the literature. Besides the traditional field of Public Policy Analysis, Operational Research (OR) has also been often used to analyse public policies and recommend actions. A relatively recent push for evidence-based management (Pfeffer and Sutton 2006), which has also been extended to public decision making (Tavakoli et al. 2000), has increased the attractiveness of formal policy analysis. In this section we briefly review such developments.

3.2.1 Public Policy Analysis

The field of Public Policy Analysis (for an overview see the classic textbook by Dunn 2012 and also Parsons 1995) has its roots in Economics and in Political Science. Economists' main focus is on how market-structures can better allocate scarce public resources, while political scientists are mostly concerned with understanding the roles of politics and the government in public policy processes (Lindblom and Woodhouse 1993).

A large body of public policy analysis is devoted to retrospective (*ex post*) analysis, which tries to understand the causes and consequences of policies after they have been implemented. Equally relevant in policy analysis is the role of prospective (*ex ante*) analysis, which encompasses the

⁴ http://ec.europa.eu/regional_policy/sources/docgener/guides/cost/guide2008_en.pdf

⁵ http://ec.europa.eu/regional_policy/sources/docgener/evaluation/evalsed/guide/index_en.htm

⁶ http://www.hm-treasury.gov.uk/d/green_book_complete.pdf;
http://www.hm-treasury.gov.uk/d/magenta_book_combined.pdf

⁷ <http://www.bis.gov.uk/assets/biscore/better-regulation/docs/i/11-1111-impact-assessment-guidance.pdf>

⁸ http://www.minambiente.it/export/sites/default/archivio/biblioteca/dsa_linee_guida_vas_1999.pdf

⁹ http://www.installationsclassees.developpementdurable.gouv.fr/IMG/pdf/Guide_PPRT_tbd_complet.pdf

¹⁰ World Bank (2010) Cost-Benefit analysis in world bank projects. Washington DC

forecasting of consequences if policies were to be implemented and prescriptions about which policies should be implemented (Dunn 2012).

Common tools for prescriptive analysis in this field are net present value assessments of costs and benefits of potential public policies, as well as cost-benefit analysis (Munger 2000), and cost effectiveness analysis, the latter often employed for military expenditure decisions (Dunn 2012). Challenges in those analyses involve how to properly monetise all benefits and how to set up an adequate discount rate for public goods without market prices (Ackerman and Heinzerling 2004; Adler and Posner 2006; Montibeller and Franco 2011).

The Evidence-Based Policy Making approach, introduced by the UK government in the 1990s, represents the most recent “practice oriented” attempt to strength the policy making process: “... *evidence-based policy helps people make well informed decisions about policy, programmes and projects by putting the best available evidence from research at the heart of policy development and implementation*” (Davies 2004). This attempt extends the idea of governing based on “facts” (instead of ideology) typical of the European culture since the enlightenment (for a discussion see Dryzek 2006). However, this approach, criticised under many aspects (see Almquist et al. 2012; De Marchi et al. 2012), failed to become a standard, perhaps the main reason was its inability to convince policy makers that the typical difficulties of legitimating public policies could be overtaken by using it.

3.2.2 Public Sector Operational Research

After the early developments of OR, and successful military applications during World War II, business applications followed, particularly in the USA. In the 1960s RAND started using OR methods for public decisions, initially only dealing with hard data, but later on developing methods for policy analysis which could take into account soft factors, such as values and subjective judgments, as well as future scenarios (RAND 1996). In the late 1960s the British central government implemented an OR group, with the intention of promoting the use of OR in decision and policy making (Kirby 2000).

Some limitations of traditional OR methods, such as an over-reliance on quantitative data and the use of an expert mode of analysis (Franco and Montibeller 2010), led to the development, in Britain, of Problem Structuring Methods (for an overview Mingers and Rosenhead 2001). These methods rely heavily on participative engagement with decision makers, adopting a facilitative mode of engagement (Franco and Montibeller 2010), and simple, often qualitative, models. Such methods have been used extensively in public sector decisions, particularly in Britain and continental Europe (Rosenhead 1992; Shaw et al. 2006).

Another important source of policy analysis support was the development of decision analysis in late 1960s (Raiffa 1968), with the use of expert judgement in defining subjective probabilities of outcomes, and further extensions to decision with multiple objectives in mid 1970s (Keeney and Raiffa 1993). Decision Analysis has been used extensively since then for the analysis of many important policy decisions, particularly in the USA (e.g. Merkhofer and Keeney 1987, Rosoff and Von Winterfeldt 2007). A more recent stream of development is the use of decision analysis embedded in recurrent processes of public prioritisation (e.g. Del Rio Vilas et al. 2013; Bana e Costa et al. 2008).

Last, but not least we should mention the literature about Data Envelopment Analysis and more generally about performance measurement in the public sector (initiated by an application about the effectiveness of public schools, see Charnes et al. 1978). (The reader can have a more comprehensive idea of the subject in Thanassoulis 1995 and Emrouznejad et al. 2008.)

3.3 Business Analytics

Business analytics has been initially developed mainly for the private sector (Davenport and Harris 2007, Davenport et al. 2010). Seen from a very pragmatic point of view, “Analytics” is an umbrella term under which many different methods and approaches developed in the past converge. These include statistics, data mining, business intelligence, knowledge engineering and extraction, decision support systems and, to some extent, operational research and decision analysis. The key idea consists in developing methods through which it is possible to obtain useful information and knowledge for some purpose, this typically being conducting a decision process in some business application.

The smart part in developing “Analytics” has been to merge different techniques and methods in order to optimise both the learning dimension as well as its applicability in real decision processes. In recent years “Analytics” has been associated to the term “Big Data” in order to take into account the availability of large data bases (and knowledge bases as well) possibly in open access (Open Data organisations are now becoming increasingly available). Such data come in very heterogeneous forms and a key challenge has been to be able to merge such different sources, in addition to solving the hard algorithmic problems presented by the huge volume of data available.

However, the mainstream approach developed in these years is based on two limiting hypotheses. The first is that the learning process is basically “data driven” with little (if any) attention paid to the values which may also drive learning. This is with regard both to “what” matters and also to “why” it matters, potentially incorporating considerations of the extent to which different stakeholder perspectives are valued or trusted. The second is that, in order to guarantee the efficiency of the learning process seen, from an algorithmic point of view, as a process of pattern recognition, it is necessary to use “learning benchmarks” against which it is possible to measure the accuracy and efficiency of the algorithms. While this perspective makes sense for many applications of machine learning it is less clear how it can be useful in cases where learning concerns values, preferences, likelihoods, beliefs and other subjectively established information, which is potentially revisable as constructive learning takes place.

4 Policy Analytics

4.1 A Definition of the Concept

To support policy makers, decision analysts need to use existing information (facts, science, local knowledge, best practices, etc.), need to constructively model opinions, values, and judgments of the stakeholders, and need to do so in a meaningful, operational and legitimate way. We denote this set of skills under the term of “policy analytics”: i.e. the use and/or development of adequate methods and suitable processes for creating purposeful knowledge, aiming at supporting those specific type of decisions processes occurring within a policy cycle.

4.2 What Can Policy Analytics Offer?

So what can Policy Analytics offer to policy makers dealing with complex policy decisions? We suggest some avenues that might be explored by decision analysts.

In terms of data requirements, we see two major roles for Policy Analytics, in the same way that there are for Business Analytics. The first one is to explore databases which already exist. The second one is to gather data and create new databases, to explore particular issues relevant to policy makers. However, the nature of policy decisions, which we described earlier, makes these two roles very distinct from Business Analytics.

In exploring data which already exists (such as citizens' votes, preferences and demands, their relation to demographics, etc.) decision analysts must recognise that the multiple stakeholders involved and/or affected by a decision, and the multiple objectives pursued by the policy makers, mean that benchmarks cannot be easily defined or arbitrarily set. Policy Analytics thus needs to focus on constructive approaches, including constructive benchmarking and learning, to support decisions in the policy cycle. The need to construct benchmarks rather than simply basing them on measured performances or behaviours, or arbitrarily defining them, requires an understanding of the stakeholders' values and their power structure and, often also calls for their engagement in the design of the analysis. The need to promote constructive learning requires a focus on models for learning rather than models to give the "right" answer (De Geus 1988) and a recognition that the role of the analytical models is a constructive one (Watzlawick et al. 1974; Roy 1993),

For example, consider a situation in which healthcare providers are assessed on the basis of the average time a patient stays in the hospital. This apparently simple indicator is subject to many contextual factors alongside differing social and political interpretations. For example, in some countries "long stays" may be considered by patients as an indicator of "ineffective" and "expensive" health services, in particular where payment is direct or through purchased insurance, as in the USA. On the other hand, in countries where healthcare is provided by the state "long stays" could be considered as high quality care, indicating a health service which does not discharge a patient until assured of a full recovery. Yet another perspective is that long stays are indicative of inadequate provision of continuing care, resulting in frustration for the patient who cannot be discharged from hospital because the appropriate support is not available in the community. We are not going to discuss whether such approaches to health are justified or not. What is important is to note that the same statistical information (data) is likely to be interpreted differently depending on context, values and culture.

Now consider the different stakeholders involved in health care management: the patients and their families, the health workers, the managers of the service and the political authorities under which health care is delivered in terms of public policy. Once again the same information (average time a patient stays in a hospital) will be considered differently depending on the stakeholder's objectives; for instance, the general manager of a hospital under pressure to increase the intake of new patients versus the patient wishing to be fully treated versus the politician wanting to be re-elected. In using this information such as this to inform any policy decision we need to consider the purpose, what is valued by whom and how, the context and culture where the policy is being developed and where it will be implemented. In order for a policy to be "legitimated" it will need to address such multiple (possibly conflicting) concerns or to be imposed by a policy maker owning a legitimation obtained beyond that precise context. In the first (and in practice more usual) case, legitimation is obtained by exchanging resources: a critical resource being information and knowledge. It is at this stage that "analytics" becomes crucial, since it should provide not just supporting information, but legitimating information¹¹ (showing for instance that reducing the average stay in hospital within a welfare state context will in the long term result in either worsening the quality of the service or in increasing its long term costs). The construction of such information requires taking into account long term effects and consequences consistent with the timing of most policy cycles. Finally it should be noted that relevance is perceived, registered (and measured) in specific moments where decisions are deliberated.

Perhaps an even more important role for Policy Analytics is its potential for creating new databases, gathering information which is relevant for the analysis and decision making, and exploring the data to support the policy-making cycle. Again constructivism (Watzlawick 1984; Roy 1993) provides a proper conceptual background for this role, in our view, given the complex nature of societal

¹¹ Supporting information is any argument supporting a claim. Legitimizing information is any argument allowing a claim to be legitimated by the participants of a discussion.

problems and the need to understand multiple perspectives, consider multiple impacts on different sectors of the society, and assess options under multiple and often conflicting objectives.

Another key aspect, is that Policy Analytics must emphasise value-driven analysis which can support value-driven decision-making (Keeney 1996), rather than being highly data-driven as is often the case in the context of business analytics. A value driven analysis understands that the alternative policies are means to achieve the values and objectives that society is pursuing. Different policies will have different impacts on the extent to which such values are achieved or upheld and, as these values often represent objectives held by different stakeholders, may impact unevenly on different segments of society.

Related to this, another opportunity which has to date been often neglected by policy analysis and public sector OR, is in supporting the design of better policies with a value-driven analysis perspective for example (Gregory and Keeney 1994; Montibeller and Franco 2011). Indeed within this perspective, the analysis is seen as supporting pro-active policy making, which tries to address problems and improve society, instead of reactively coping with public dissatisfaction and complaints.

We envisage several opportunities for the use of Policy Analytics in organisational contexts. A major opportunity is to embed analyses to inform decisions in the policy analytic cycle, for instance, in helping the identification of issues, in predicting impacts of possible policies, in policy design, in simulating policy implementation, and in helping the evaluation and monitoring of implemented policies. In Table 1 we suggest how Policy Analytics can support each step in the policy cycle and try to distinguish it from how Business Analytics would provide support in a similar business setting.

5 Conclusions and Further Directions of Development

In this paper we have suggested a conceptualisation for policy analytics and proposed a framework for its use in supporting public decisions processes occurring within a policy cycle. We argued that there is demand for policy analysis, and reviewed briefly methods that provide formal analysis in this context, such as public policy analysis, public sector operational research, decision analysis, and data development analysis applied to assessing public organisations.

We also argued that business analytics, which was developed mainly for supporting decisions in the private sector, while powerful and increasingly applied, has two limiting hypotheses: it is data driven and needs clear benchmarks to be set. We believe that these are challenging if one wants to apply analytics in supporting the policy cycle.

Let us now discuss the three motivations for which we started this discussion:

- *Are practitioners involved in supporting decision processes in the public domain appropriately using analytics and, if so, what type of analytics could fit the type of support they use?* Only to some extent. First, the types of analytics readily available have been designed for business purposes. Despite being generally helpful (in allowing data to support more effectively decision processes), business analytics may not fully fit the requirements of the policy cycle as these have been presented in this paper. Second, policy making is, and will remain, essentially a value driven decision process. Besides learning from data we need to learn from values: under such a perspective practitioners need tools, methods, and models allowing the consideration of this perspective. Furthermore, the clients (policy makers) need also to understand the difference between data versus value-driven analysis: if offering simple data driven decision support can be misleading, demanding simple data driven policy support can also be misleading. Policies are not in the data, they are in the values. Third, such tools, methods, and models need to support the whole policy cycle.

- *Is decision analysis training appropriate for helping the integration of traditional decision analytic tools and analytics in the area of public policy making?* We think that perhaps not in full. Most of the training of decision analysts does not address the specific requirements of the policy cycle, while most of the training of policy scientists does not address the issue of using formal methods of decision support. We need to establish training for decision analysts that could cover the whole spectrum of policy cycle issues, potentially from an interdisciplinary perspective and incorporating awareness of the social and political contexts of policy making. In doing so we must address decision support as a methodology, not just as a collection of tools and methods. We need decision analysts able to understand the complexity of the policy cycle, to be flexible enough to shift from one tool to another one (from problem structuring methods to quantitative modelling, from learning procedures to justification construction), to be aware of the possibilities offered by the new technologies but also to be clever enough to construct new paths within it.

- *Are researchers giving appropriate consideration to the relationships between the specificities of public policy making and analytics?* The above discussion introduces a number of challenges for researchers. If we think that such challenges need to be pursued, then our research needs to address them. In the following we present an agenda of issues we consider relevant (but not it is not an exhaustive list).
 - *Preference Learning.* If policy making should reflect societal values, then we need to learn about them (the policy makers' values, stakeholders' values, etc.). Values are often operationalized as preference statements, either comparative or absolute ones. The field of preference learning is increasingly becoming an important research area (see Fürnkranz and Hüllermeier 2010) addressing such issues, although more from a machine learning perspective; Belton et al. (2009) discuss the interaction between individual learning and model learning and associated research challenges in the context of interactive decision making. A more constructive learning approach needs to be developed (see Bruner 1986; Mousseau, 2005).
 - *Scenario Planning.* The long-term implications of policy making imply the need to consider the range of possible futures, sometimes characterised by deep uncertainties and calling for the development of future scenarios (see Godet 2000; Montibeller et al. 2006; Schroeder and Lambert 2011; Ram and Montibeller 2012; Stewart et al. 2013). Further research on how scenarios are constructed (from paths along a decision tree to precise configurations of interaction spaces or arenas) and how to address issues of robustness in scenario planning would be welcome (see Roy 1998; Vincke 1999; Levy et al. 2000; Wong and Rosenhead 2000; Perny and Spanjaard 2003; Liesö et al. 2007).
 - *Argumentation Theory.* As already introduced by Habermas (1981), legitimated policies are the ones which are appropriately explained, justified, supported and not sufficiently confuted (i.e. argued). Argumentation theory (Aristotle 1990; Schopenhauer 1864; Toulmin 1958) establishes a formal and rational framework for how to construct, use, exchange and confute arguments. Although hardly used as support in policy analysis (for exceptions see Atkinson et al. 2004; Rehg et al. 2005; Cartwright and Atkinson 2008; Modgil and Prakken 2012) there is scope for further investigating several of its dimensions: the construction of explanations and justifications; the construction of argumentation scenarios; the issue of legitimate arguments (in collective debate); the relationship between game theory and argumentation theory.
 - *Support for Problem Structuring and Formulation.* A large part of the decision support activities occurring within a policy cycle are about understanding, formulating and structuring "problems". Problem structuring methods (see Franco et al. 2006; Shaw et al. 2007) are now widely acknowledged as part of decision analytic tools; and there is a growing but still small body of research and practice on how to integrate such methods with other formal and/or

quantitative methods (for examples see Belton et al. 1997; Bana e Costa et al. 1999; Montibeller and Belton 2006; Montibeller et al. 2007a; Montibeller et al. 2007b; Belton and Stewart 2010; and Howick and Ackermann 2011 for a survey of practice in OR). More importantly, from an analytics point of view, there are challenges on how problem formulation and problem structuring can be conducted in the absence of small-groups of decision makers, which has been typically the case in traditional applications of problem-structuring methods.

- *Reformulation of Decision Problems.* Policy cycles involve, often, long decision processes. They also are “learning processes” for stakeholders implied in the cycle. Updating, revising and reformulating decision models is a regular activity within policy cycles and are all activities which could potentially be supported. This is an issue already addressed in the literature, both from a general point of view (see Gärdenfors 1988; Katsuno and Mendelzon 1991) and for specific decision support purposes (see Tsoukiàs 1991; Liberti 2009), but we are still far from having a comprehensive framework which could be also practically applied (although we have now specific formal languages for some classes of reformulation problems, see Liberti et al. 2010).
- *Design of Alternatives.* Most decision problems discussed in the literature consider the set of alternatives on which they apply as “given”, although we know that in practice frequently such a set needs to be constructed. There is little in the literature addressing this problem (see Belton and Stewart 2002 for a brief overview), despite the awareness of it (for example, Keeney 1996; Goodwin and Wright 1998; Keller and Ho 1988). However, policy makers rarely come with established alternatives. Most of the policy cycle is about designing or constructing alternatives. Actually, most of “smart” policy making is about “innovative design” of “innovation policies”, that (Montibeller and Franco 2011) is designing alternatives considered unconceivable at that moment in time (creative design). Simon (1954) had already discussed this cognitive activity in his seminal work, without providing operational and/or formal methods for addressing it. More recently Hatchuel and Weil (2009) introduced C-K theory (C-K = concept-knowledge) as a general design theory opening a way to extending decision analysis by operationally addressing this issue which merits further exploration. There have also been suggestions for value-focused brainstormings of decision alternatives (Keeney 2012; Montibeller et al. 2009), an approach which is resonant with Corner et al. (2001) dynamic decision problem structuring.
- *Decision Aiding Practice.* The great majority of research in our field concerns “theory”. As emphasised in Tsoukiàs (2007) it focuses on how decisions are “taken”, underestimating both the theoretical and practical problems of how to support decision processes. Moreover, while the issue of introducing decision models within an organisational context has been discussed in the literature (Nutt 1993), very little has been done as far as inter-organisational contexts are concerned (see Munda 2008). Given the specificities of the policy cycle, exploring the practical aspects of providing decision support remains a key research issue.
- *Interdisciplinary research.* The above mentioned research priorities clearly call for more interdisciplinary research engaging disciplines such as Artificial Intelligence, Computer Science, Sociology, Policy Analysis, Decision Analysis, and Cognitive Sciences. The reader will note that the references list of this paper already cites contributions from all such areas. Either Policy Analytics will emerge as a strong interdisciplinary research area or it will never succeed.

Concluding, we hope to have highlighted that Policy Analytics represents a key opportunity for the future of Decision Analysis, but poses challenges with implications for research, training and practice. Perhaps most importantly, it may provide an opportunity for how the world around us can be improved by decision analysts.

References

- Ackerman F, Heinzerling L (2004) *Priceless: On Knowing the Price of Everything and the Value of Nothing*. The New Press
- Adler MD, Posner ED (2006) *New Foundations of Cost Benefit Analysis*. Harvard University Press
- Almquist R, Grossi G, Jan van Helden G, Reichard Ch (2012) Public sector governance and accountability. Critical perspectives of accounting, in press. doi:10.1016/j.cpa.2012.11.005
- Aristotle (1990) *Nicomachean Ethics*. Oxford University Press, Oxford, Originally published in 350bc, english edition by I. Bywater.
- Atkinson K, Bench-Capon T, Mcburney P (2004) *Parmenides: Facilitating democratic debate*. Electronic Government, LNCS 3183, Springer Verlag, Berlin 313-316
- Bana e Costa CA, Ensslin L, Correa EC, Vansnick J-C (1999) Decision Support Systems in action: Integrated application in a multicriteria decision aid process. *European Journal of Operational Research* 113:315-335
- Bana e Costa CA, Lourenço JC, Chagas MP, Bana e Costa JC (2008) Development of Reusable Bid Evaluation Models for the Portuguese Electric Transmission Company. *Decision Analysis*, 5:22-42
- Belton V, Stewart TJ (2002) *Multiple Criteria Decision Analysis: An Integrated Approach*. Kluwer Academic, Dordrecht
- Belton V, Ackermann F, Shepherd I (1997) "Integrated support from problem structuring through to alternative evaluation using COPE and V•I•S•A". *Journal of Multiple Criteria Analysis* 6:115-130
- Belton V, Stewart TJ (2010) Problem Structuring for Multiple Criteria Analysis, in *New Trends in Multicriteria Decision Analysis*. In Greco S, Figueira RJ, Ehrgott M (Ed) Springer International Series in Operational Research and Management Science, pp. 209-240
- Belton V, Branke J, Eskelinen P, Greco S, Molina J, Ruiz F, Slowinski R (2009) Interactive Multiobjective Optimization from a Learning Perspective, in *Multiobjective Optimization – Interactive and Evolutionary Approaches*. In Branke J, Deb K, Miettinen K, Slowinski R (Ed) LNCS State-of-the-Art Surveys (5252) Springer, pp. 405-434.
- Bouyssou D, Marchant Th, Pirlot M, Perny P, Tsoukiàs A, Vincke Ph (2000) *Evaluation and Decision Models: a critical perspective*. Kluwer Academic, Dordrecht
- Bruner J (1986) *Actual Minds, Possible Worlds*. Cambridge, MA, Harvard University Press
- Cartwright D, Atkinson K (2008) Political Engagement Through Tools for Argumentation. *Proceedings of COMMA*, 116-127.
- Charnes A, Cooper W, Rhodes E (1978) Measuring the efficiency of decision-making units. *European Journal of Operational Research* 2:429-444.
- Corner J, Buchanan J, Henig M (2001) Dynamic decision problem structuring. *Journal of Multi-Criteria Decision Analysis* 10:129-141
- Dasgupta P, Pearce DW (1972) *Cost-benefit analysis: Theory and practice*. Macmillan, Basingstoke

- Davenport TH, Harris JG (2007) *Competing on Analytics: The New Science of Winning*. Harvard Business Press
- Davenport TH, Harris JG, Morison R (2010) *Analytics at Work: Smarter Decisions, Better Results*. Harvard Business Press
- Davies PT (2004) Is evidence-based government possible? Jerry Lee Lecture. <http://www.nationalschool.gov.uk/policyhub/downloads/JerryLeeLecture1202041.pdf>
- De Marchi G, Lucertini G, Tsoukiàs A (2012) From Evidence Based Policy Making to Policy Analytics. *Cahier du LAMSADE 319*, Université Paris Dauphine
- Dollery B, Worthington A (1996) The Evaluation of Public Policy: Normative Economic Theories of Government Failure. *Journal of Interdisciplinary Economics* 7:27-39
- Dorling D, Simpson S (1999) *Statistics in Society*. Hodder Arnold, London
- Dryzek JS (2006) Policy Analysis as critique. In Moran M, Rein M, Goodin RE (Ed.) *The Oxford Handbook of Public Policy*. Oxford University Press, Oxford, pp. 190-203
- Dunn WN (2012) *Public Policy Analysis*, 5th edn. Pearson
- Emrouznejad A, Barnett RP, Tavares G (2008) Evaluation of research in efficiency and productivity: A survey and analysis of the first 30 years of scholarly literature in DEA. *Socio-Economic Planning Sciences* 42:151-157
- Franco LA, Shaw D, Westcombe M (2006) Special Issue: Problem Structuring Methods I. *Journal of the Operational Research Society* 57:757-883
- Franco LA, Montibeller G (2010) Facilitated modelling in operational research. *European Journal of Operational Research*, 205:489–500
- Fürnkranz J, Hüllermeier E (2010) *Preference Learning*. Springer Verlag, Berlin
- Gärdenfors P (1988) *Knowledge in flux*. MIT Press, Cambridge
- De Geus AP (1988) Planning as learning. *Harvard Business Review* 66:70–74
- Godet M (2000) *The art of scenarios and strategic planning: tools and pitfalls*. Elsevier Science Inc., New York
- Goodwin P, Wright G (1998) *Decision Analysis for Management Judgment*. J. Wiley, New York
- Gregory RS, Keeney RL (1994) Creating Policy Alternatives Using Stakeholder Values. *Management Science* 40:1035–1048.
- Habermas J (1981) *The Theory of Communicative Action*. Translated (1984-87) by Th. McCarthy, Cambridge: Polity
- Habermas J (1990) *Logic of the Social Sciences*. MIT Press, Boston
- Hatchuel A, Weil B (2009) C-K design theory: An advanced formulation. *Research in Engineering Design* 19:181–192
- Hill M (1997) *The Public Policy Process*. Pearson Education Limited, Harlow, England
- Howick S, Ackermann F (2011) Mixing OR methods in practice: past, present and future directions. *European Journal of Operational Research* 215:503–511

- Juntti M, Russel D, Turnpenny J (2009) Evidence, politics and power in public policy for the environment. *Environmental Science and Policy* 12:207-215
- Katsuno H, Mendelzon AO (1991) On the difference between updating a knowledge base and revising it. *Proceedings of KR'91*, 387-394.
- Keeney RL (2012) Value-Focused Brainstorming. *Decision Analysis* (October 8). doi:10.1287/deca.1120.0251. <http://da.journal.informs.org/content/early/2012/10/05/deca.1120.0251>.
- Keeney RL (1996) *Value-Focused Thinking: A Path to Creative Decisionmaking*. Harvard University Press
- Keeney RL, Raiffa H (1993) *Decisions with multiple objectives: preferences and value tradeoffs*. 2nd edn., Cambridge University Press, Cambridge
- Keller LR, Ho JL (1988) Decision Problem Structuring: Generating Options. *IEEE Transactions on Systems, Man, and Cybernetics* 18:715-728
- Kirby MW (2000) Operations Research Trajectories: The Anglo-American Experience from the 1940s to the 1990s. *Operations Research*, 48:661-670
- Kraft M, Furlong SR (2007) *Public Policy. Politics, Analysis and Alternatives*. 2nd edn. CQ Press, Washington
- Larson RC, Odoni AR (1981) *Urban Operations Research*. Prentice Hall, NJ
- Lasswell HD (1956) *The Decision Process: Seven Categories of Functional Analysis*. University of Maryland Press, College Park
- Levy JK, Hipel KW, Kilgour DM (2000) Using environmental indicators to quantify the robustness of policy alternatives to uncertainty. *Ecological Modelling* 130:79-86
- Liberatore M, Luo W (2010) The Analytics Movement: Implications for Operations Research. *Interfaces* 40:313-324
- Liberti L (2009) Reformulations in mathematical programming: definitions and systematics. *RAIRO* 43:55-85
- Liberti L, Cafieri S, Savourey D (2010) The Reformulation-Optimization Software Engine. *Proceedings of ICMS* 303-314
- Liesiö J, Mild P, Salo A (2007) Preference Programming for Robust Portfolio Modeling and Project Selection. *European Journal of Operational Research* 181:1488 -1505
- Lindblom CE, Woodhouse EJ (1993) *The Policy-Making Process*. 3rd edn., Prentice Hall
- Merkhofer MW, Keeney RL (1987) A Multiattribute Utility Analysis of Alternative Sites for the Disposal of Nuclear Waste. *Risk Analysis* 7:173-194
- Mingers J, Rosenhead J (2001) *Rational analysis for a problematic world revisited: problem structuring methods for complexity, uncertainty and conflict*. Wiley
- Modgil S, Prakken H (2012) A general account of argumentation with preferences. *Artificial Intelligence*, <http://dx.doi.org/10.1016/j.artint.2012.10.008>
- Montibeller G, Belton V (2006) Causal Maps and the Evaluation of Decision Options. *Journal of the Operational Research Society* 57:779-791

- Montibeller G, Belton V, Lima MVA (2007a) Supporting factoring transactions in Brazil using reasoning maps: a language-based DSS for evaluating accounts receivable. *Decision Support Systems* 42:2085-2092
- Montibeller G, Belton V, Ackermann F, Ensslin L (2007b) Reasoning Maps for Decision Aid: An Integrated Approach for Problem-structuring and Multi-criteria Evaluation. *Journal of the Operational Research Society* 59:575–589
- Montibeller G, Gummer H, Tumidei D (2006) Combining scenario planning and multi-criteria decision analysis in practice. *Journal of Multi-Criteria Decision Analysis* 14:5-20
- Montibeller G, Franco LA, Lord E, Iglesias A (2009) Structuring Resource Allocation Decisions: A Framework for Building Multi-criteria Portfolio Models with Area-grouped Options. *European Journal of Operational Research* 199:846-856
- Montibeller G, Franco LA (2011) Raising the bar: strategic multi-criteria decision analysis. *Journal of the Operational Research Society* 62:855-867
- Moran M, Rein M, Goodin GE (2006) *The Oxford Handbook of Public Policy*. Oxford University Press, Oxford.
- Mousseau V (2005) A general framework for constructive learning preference elicitation in multiple criteria decision aid. *Cahier du Lamsade* 223, Université Paris Dauphine
- Munda G (2008) *Social Multi-Criteria Evaluation for a Sustainable Economy*. Springer Verlag, Berlin
- Munger MC (2000) *Analyzing policy: choices, conflicts, and practices*. WW Norton
- Nas TF (1996) *Cost-benefit analysis: Theory and application*. Sage Publications
- Nutley SM, Walter I, Davies HTO (2003) From knowing to doing: A framework for understanding the evidence-into-practice agenda. *Evaluation* 2:125-148
- Nutt PC (1993) The formulation processes and tactics used in organizational decision making. *Organization Science* 4:226-251
- Ostanello A, Tsoukiàs A (1993) An explicative model of 'public' interorganisational interactions. *European Journal of Operational Research* 70:67-82
- Ostrom E (1986) An agenda for the study of institutions. *Public Choice* 48:3-25
- Parsons DW (1995) *Public policy: an introduction to the theory and practice of policy analysis*. Edward Elgar
- Perny P, Spanjaard O (2003) An Axiomatic Approach to Robustness in Search Problems with Multiple Scenarios. *Proceedings of UAI03*, 469-476
- Pfeffer J, Sutton RI (2006) *Hard Facts, Dangerous Half-Truths and Total Nonsense: Profiting from Evidence-based Management*. Harvard Business Press
- Pollock SM, Rothkopf MH, Barnett A (1994) *Operational Research and the Public Sector*. North Holland, Amsterdam.
- Raiffa H (1968) *Decision analysis: introductory lectures on choices under uncertainty*. Addison-Wesley England, Oxford
- Ram C, Montibeller G (2012) Exploring the impact of evaluating strategic options in a scenario-based multi-criteria framework. *Technological Forecasting & Social Change*, Forthcoming

- Rehg W, McBurney P, Parsons S (2005) Computer decision-support systems for public argumentation: assessing deliberative legitimacy. *AI & Society* 19:203-228
- Del Rio Vilas VJ, Voller F, Montibeller G, Franco LA, Sribashayam S, Watson E, Hartley M, Gibbens JC (2013) An integrated process and management tools for ranking multiple emerging threats to animal health. *Preventive Veterinary Medicine* 108:94-102
- RAND (1996) 50th Project Air Force. RAND Corporation. Available at: <http://www.rand.org/content/dam/rand/www/external/publications/PAFbook.pdf>
- Rosenhead J (1981) Operational Research in Urban Planning. *Omega* 9:345-364
- Rosenhead J (1992) Into the Swamp: The Analysis of Social Issues. *The Journal of the Operational Research Society* 43:293
- Rosoff H, Von Winterfeldt D (2007) A Risk and Economic Analysis of Dirty Bomb Attacks on the Ports of Los Angeles and Long Beach. *Risk Analysis* 27:533-546
- Roy B (1993) Decision science or decision-aid science? *European Journal of Operational Research* 66:184-203.
- Roy B (1998) A missing link in {OR-DA}: Robustness analysis. *Foundations of Computing and Decision Sciences* 23:141-160
- Schroeder MJ, Lambert JH (2011) Scenario-based multiple criteria analysis for infrastructure policy impacts and planning. *Journal of Risk Research* 14:191-214
- Schopenhauer A (1864) *Dialektik in Schopenhauers handschriftlicher Nachlaß*. J. Frauenstädt (ed.), Brockhaus, Leipzig
- Shapiro S, Morrall JF (2012) The Triumph of Regulatory Politics: Benefit-Cost Analysis and Political Salience. *Regulation and Governance* 6:189-206
- Shaw D, Franco LA, Westcombe M (2006) Problem structuring methods: new directions in a problematic world. *Journal of the Operational Research Society* 57:757-758
- Shaw D, Franco LA, Westcombe M (2007) Special Issue: Problem Structuring Methods II. *Journal of the Operational Research Society* 58:545-700
- Simon HA (1954) A behavioral model of rational choice. *Quarterly Journal of economics* 69:99-118
- Smit HTJ, Trigeorgis L (2004) *Strategic Investment: Real Options and Games*. Princeton University Press
- Stewart ThJ, French S, Rios J (2013) Integrating multicriteria decision analysis and scenario planning: Review and extension, *Omega* 41:679-688
- Tavakoli M, Davies HTO, Thomson R (2000) Decision analysis in evidence-based decision making. *Journal of Evaluation in Clinical Practice* 6:111-120
- Thanassoulis E (1995) Assessing police forces in England and Wales using data envelopment analysis. *European Journal of Operational Research* 87:641-657
- Toulmin S (1958) *The Uses of Argument*. Cambridge University Press, Cambridge
- Trigeorgis L (1990) A Real Options Application in Natural Resource Investments. *Advances in Futures and Options Research* 4:153-164

- Tsoukiàs A (1991) Preference Modelling as a Reasoning Process: a new way to face Uncertainty in Multiple Criteria Decision Support Systems'. *European Journal of Operational Research* 55:309-318
- Tsoukiàs A (2007) On the concept of Decision Aiding Process. *Annals of Operations Research* 154:3-27
- Tsoukiàs A (2008) From Decision Theory to Decision Aiding Methodology. *European Journal of Operational Research* 187:138-161
- Vincke Ph (1999) Robust solutions and methods in decision aid. *Journal of Multi-Criteria Decision Analysis* 8:181-187
- Watzlawick P (1984) *The invented reality: how do we know what we believe we know?* Norton, New York
- Watzlawick P, Weakland JH, Fisch R (1974) *Change; principles of problem formation and problem resolution.* Norton, New York
- White L, Bourne H (2007) Voices and values: Linking values with participation in OR/MS in public policy making. *Omega* 35:588-603
- Wong H-Y, Rosenhead J (2000) A rigorous definition of robustness analysis. *Journal of the Operational Research Society* 51:176-182
- World Bank (2010) Cost-Benefit analysis in world bank projects. Washington, DC. Available at: http://siteresources.worldbank.org/INTOED/Resources/cba_overview.pdf

Table 1. The Role of Policy Analytics in the Policy Cycle

Steps in the Policy Cycle	Business Analytics	Policy Analytics
Issue Identification	Definition of issue by the analyst	Analyst understands perspectives from different stakeholders
Defining Policy Objectives	Data-driven definition of attributes	Value-, cultural- and stakeholder-driven definition of objectives
Policy Design	Data-driven design of alternative policies	Innovative and value-driven design of alternative policies
Policy Testing	Data-based testing and learning (data mining, predictive analysis)	Multiple tests to assess potential impacts (citizens surveys, data mining, prospective analysis, etc.)
Policy Finalisation	Sensitivity analysis of the results, given the input parameters	Robust analysis of the results, given broad issues and multiple values being considered
Policy Implementation	Implementation is typically straightforward, given the issue considered	Analysis helps implementation, mapping resistances and side effects of the policy
Policy Monitoring & Evaluation	Evaluation conducted against the success criterion initially set	Multiple and contested success criteria; evaluation is value and stakeholder based.
Policy Readjustment & Innovation	Innovation is data-driven and thus reactive	Innovation is value-driven and thus proactive