Assessing Future Water Availability in Arid Regions Using Composition and Salience of Decision Criteria

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Paper submitted for CESUN 2014 - Fourth International Engineering Systems Symposium to be held June 8-11, 2014 at Stevens Institute of Technology.
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Abstract— Water resources development options are usually selected on a least-cost basis. While economic considerations are dominant in choosing projects, there are also a mix of other factors including social demands, political expediency, social equity, and environmental considerations that impact final decisions and development of water supply systems. Understanding local priorities in water resource management decisions can allow for forming expectations of future regional water availability. In this research, we propose that future water availability in arid regions may be assessed by considering key projects that have been identified or planned by regional experts. Using Multi-Criteria Decision Analysis methods as a framework to organize set of decision criteria and their relative salience, the likelihood of selection (and development) of a project can be determined and used to form expectations of future regional water availability. We use this approach in a case study for Jordan, and find that large-scale desalination projects – that have been in the planning books for decades - are now most likely to be pursued and implemented in the country. Finally, we discuss strengths, limitations, and the general applicability of this method for assessing future water availability in other arid regions.

Keywords—multi-criteria decision analysis; decision theory; water supply planning; water management; infrastructure development

I. INTRODUCTION

Demands for fresh water continue to grow with increasing population, and competition across different economic sectors has intensified for this constrained resource [1]. Public sector planners as well as private sector developers, making investment decisions for infrastructure expected to function for decades into the future increasingly need to assess mid-term and long-term water availability. In general, it is difficult to make future assessments of availability, and we cannot fully rely on historical trends for forecasting. Despite the challenges, however, it is still worthwhile to develop and use methods that can allow for forming reasoned expectations.

In this work, we approach the question of estimating future availability by analyzing projects under consideration by regional planners, and assessing the likelihood of their implementation and the consequent impact on local water supply. We consider ‘availability’ as water supply that is controlled by decision makers and engineered projects – we are not modeling water resources from a hydro-geological standpoint, or the natural hydro-climatic system of a region.

In order to assess the likelihood of implementation of projects, we follow a quantitative and qualitative empirical methodology in which we determine criteria employed by key decision-makers, and use that information to form an assessment of which projects are likely to be chosen. We use Multi-Criteria Decision Analysis (MCDA) to elicit and encode decision factors and their relative importance, and then analyze the full combinatorial set of options and preferences to quantify likelihood of implementation of future projects. This approach allows us to identify likely regional infrastructure development trajectories and consequent water supply volume that would become available over time.

II. BACKGROUND AND LITERATURE REVIEW

A variety of Multi-Criteria Decision Methods (MCDM) have been developed and used in water resources planning [2]. The methodologies range from simple scoring techniques to compare multiple alternatives to multi-objective optimization methods that include goal programming, compromise programming, and stochastic optimization among others. In water sector planning, the methods have been applied in resource management [3], channel improvements [4], long distance water transfer projects [5], and desalination project evaluations [6,7] among others.

There are two broad categories within MCDM that have been applied in water resources planning. The first involves discrete decision spaces (i.e. a set of discrete alternatives) in which the best option is to be determined given a set of criteria, and the other relates to resource allocation problems (e.g. allocating water volume among competing users) to optimize a set of objectives. In most of the studies focused on discrete alternatives evaluation, researchers have used expert and/or stakeholder interviews to determine weights for decision criteria that include economic, social, public health, technical and sustainability factors [4, 6-9]. Additive weighted sum methods have been developed and used to identify best options [10]. A key limitation of these methods, however, is...
the subjectivity, and a decision maker’s judgment influences the outcome for the ‘best’ identified alternative. Despite this limitation, these approaches are understood and used in practice. Often some form of sensitivity analysis is usually conducted to determine the impact of changing the values of the criteria and weighting coefficients to assess the robustness of the optimal result.

In this work, we identify composition and salience of decision factors by interviewing decision makers in a region for water resources development, and then use the relative performance of projects (under consideration for development) on the set of identified factors to form expectations of future supply. In other words, we use MCDA as an organizing framework to form expectations of future development decisions by using expected project performance as a proxy of likelihood of project selection. In traditional analysis, MCDA methods are employed to select a best alternative from a given set of options. In our work, we do not perform an optimization, or select an optimal solution; rather we use MCDA to create a traceable and systematic basis for forming expectations for the future. Our focus is prospective, rather than prescriptive in nature. To the best of our knowledge, MCDA methods have not been previously employed in this manner, and our work develops a new approach. Furthermore, our case study is based on real projects (as opposed to hypothetical options) and we compare our results with actual decisions that have been made to identify the strengths and limitations of our approach.

III. METHODOLOGY AND APPROACH

We base our methodology on a series of assumptions: we assume that resource development is through a discrete set of infrastructure projects. Typically, resource availability is managed through supply augmentation as well as demand management. We propose that demand management schemes can also be represented as ‘projects’ with associated costs and impacts on water volume that becomes available in the system for use as a result of demand management measures.

We also assume that the resource will be needed for direct use and cannot be substituted through trade and imports. The concept of virtual water, and embedded water in imported goods has been extensively studied for understanding how trade helps alleviate water demands and local supply shortages [11]. We do not consider these aspects in our analysis.

Additionally, we assume that a few key actors ultimately drive decisions that lead to implementation of projects, and that decisions are based on a discrete set of factors that collectively shape final choices.

Starting with these assumptions, we formulate the following steps:

1. Identify the decision-makers (DMs) and the decision space - i.e. set of projects being considered for development. This step involves stakeholder analysis for identification of key institutions, organizations, and actors who are involved in planning.

2. Using results of the stakeholders analysis, conduct discussions and surveys to determine decision criteria, and their relative importance as viewed by key decision makers. This information will typically be collected through interviews and complemented with published reports and literature.

3. Organize data of the projects and their performance on decision criteria (factors) in a performance matrix, and compare projects (alternatives) using techniques such as normalized additive weighting.

4. Compare results of analysis with past decisions (as represented by implemented and commissioned projects). This can be done by using data of projects that have been implemented, and checking how the results of the analysis match with actual decisions (and development).

5. Assess likelihood of future project selection by evaluating performance on different preference (weights) sets, and conduct a stochastic evaluation in which probabilities of importance of different criteria are varied. The results can be further analyzed to determine factors that may drive selection of particular projects in the future as well as the conditions that would lead to new salience of those factors. This allows for obtaining insights on what future scenarios may drive progression along possible different development pathways.

The details of the approach we used are as follows: Suppose $n$ projects are being considered for development. A set of $m$ criteria is employed by decision makers for evaluating these projects. An $m \times n$ performance matrix, $A$ is then defined as:

$$ A = [a_{ik}], \quad (1) $$

where $a_{ik}$ is the value (or performance) of $k^{th}$ project (or alternative) for $i^{th}$ criterion.

The performance across each criterion is normalized, and an $m \times n$ matrix, $\bar{A}$ is obtained. For criteria for which performance is desired to be minimized, $\bar{a}_{ik}$ is given as:

$$ \bar{a}_{ik} = \frac{\max(a_{ik}) - a_{ik}}{\max(a_{ik}) - \min(a_{ik})} \quad (2) $$

where $\max(a_{ik})$ denotes maximum value for $i^{th}$ criteria, $\min(a_{ik})$ denotes minimum value for $i^{th}$ criteria across all $k$ (i.e. set of projects). For criteria for which performance is desired to be maximized (such as cost), the $\bar{a}_{ik}$ is given as:

$$ \bar{a}_{ik} = \frac{a_{ik} - \min(a_{ik})}{\max(a_{ik}) - \min(a_{ik})} \quad (3) $$

A weighted, normalized vector $J$ is now defined as:
where $w = [w_1 \ldots w_m]$, with $w_i$ a weighting factor of $i$th criterion and

$$\sum_{i=1}^{m} w_i = 1. \quad (5)$$

Note that $J$ is a row vector, and consists of weighted, normalized performance values of the $n$ projects.

We next explore the combinatorial space for $w$ where $w_i$ are systematically varied from 0 to 1 with step $dW$, and selecting sets of $w$ where Eq. 5 holds. For $q$ such combinations of $w$, we determine number of cases when the $i$th project is determined to be $i^*$, where

$$i^* = \arg \max (J) \quad (6)$$

i.e. the project with highest normalized weighted performance.

If $i$th project is determined to be $i^*$ in $r$ cases (of the $q$ possible combinations of $w$), we consider the likelihood of project $i$ being selected and developed as:

$$\pi_i = \frac{r}{q}. \quad (7)$$

Thus, if project $i$ has associated annual supply volume $v_i$, then there is $\pi_i$ likelihood of $v_i$ being added to regional water availability.

Eq. 7 applies when all possible (and valid) combinations of $w$ are equally likely to occur. In reality, certain values for $w$ (that reflect particular priorities of local decision makers) should be more carefully studied, perhaps using non-uniform probability distribution functions, and it should be considered that the priorities could change based on local political, economic, and social issues.

We can describe this problem with the decision tree in Fig. 1. The branches emanating from the decision node represent options (projects) being considered. The diagram shows that for each of the $n$ projects being considered for development decision, there is a chance that various scenarios that change the context of project selection decisions may occur, wherein a particular set of decision preferences will hold. The particular set of preferences ($w_k$) will impact the performance of a given project (i.e. its $J$ is a function of $w_k$).

We define a matrix $W$, with $s$ rows where each row constitutes a scenario, a vector $w$ with values representing scenarios with different decision priorities:

$$W = \begin{bmatrix} w_{11} & \cdots & w_{1m} \\ \vdots & \ddots & \vdots \\ w_{s1} & \cdots & w_{sm} \end{bmatrix}. \quad (8)$$

If the $k$th scenario has probability $p_k$ of occurring, and vector $p$ is defined as: $p = [p_1 \ldots p_k \ldots p_s]$, where

$$\sum_{k=1}^{s} p_k = 1 \quad (9)$$

Then the expected performance is given by:

$$\tilde{J}_{[s 	imes n]} = p [1 \times s] A_{[s \times m]} \bar{A}_{[m \times n]} \quad (10)$$

The $i$th element of $\tilde{J}$, $\tilde{J}_i$, is the expected performance of project $i$, given the probability distribution $p$ and set of decision priorities $W$. The project $i^*$ will likely be selected where $i^* = \arg \max (\tilde{J})$.

IV. APPLICATION: ASSESSING FUTURE WATER AVAILABILITY IN JORDAN

Jordan is a water scarce country, with annual per capita water resources of 145 m$^3$ that are well below the scarcity threshold of 1000 m$^3$ [12]. The water resources in Jordan have been extensively studied, and a number of studies have employed MCDA methods to evaluate supply augmentation options (including desalination, long distance pumped groundwater pipelines), and demand management options (such as changing cropping patterns in the agricultural sector) [6-7, 13].

In one study, stochastic linear programming was used to plan for water supply in the capital city, Amman (host to 40% of the country’s population), under different climate change scenarios. The results recommended delaying use of water from the Disi aquifer (a trans-boundary non-renewable aquifer on the border of Jordan and Saudi Arabia) until 2060 and beginning to use desalinated water from the Red Sea in 2085 [14]. In another study, using stochastic mixed integer
optimization, a systematic evaluation of both conservation actions with new supply or loss reduction alternatives for residential and commercial water use was conducted [15]. The results showed that large megaprojects such as the Red-Dead canal (in which water from the Red Sea is conveyed to the Dead Sea) are not ideal given the availability of cheaper alternatives such as a reduction of leakages in water delivery networks.

While Jordan’s natural water resources and infrastructure options are well understood [23], recent political crises, infrastructure development for harnessing local energy resources, and longer-term climate change projections lend increased importance to the question of understanding future water availability and access in the region. The political unrest in neighboring countries has caused a large refugee migration that has burdened water supplies [13], plans for exploiting local oil-shale reserves require water for mining [17-18], and climate change projections indicate further worsening of fresh water availability in the country.

Within this context we conducted a study assessing future water availability in the country, using the methodology described in section III. We first performed a detailed assessment of the stakeholders in the water sector in Jordan, and conducted more than 30 interviews across a range of stakeholder groups from highest-ranking decision makers (ministers and secretary generals) to rural community users and farmers. Table 1 provides a summary of the organizations and actors who where interviewed in two field trips to the country in 2012 and 2013.

Table 2 shows that cost was consistently one of the top two driving factors listed by the respondents. It is also useful to note that most DMs added their own factors that included ‘geographic distribution’ (as a proxy for access equity), ‘sectoral / social priorities’, and ‘sustainability of supply’. In Jordan, where some water resources are non-renewable, the sustainability of the supply source (in terms of annual renewable water supply) is an important factor.

In the next step, we identified the decision space by compiling information regarding major water infrastructure projects under consideration using agencies reports [12], and included only large-scale projects that are likely to have the greatest impact on total water availability in the country (see Table III). These projects collectively have the largest impact on water supply volume, however the list in Table III is not exhaustive, and there are a number of small-scale community

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<th>Societal Demand</th>
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<td>6</td>
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a Ministry of Water and Irrigation
b Ministry of Planning and International Cooperation
c Ministry of Environment

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<td>Former Ministers, Secretary Generals, Directors of Planning</td>
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<td>Project Management Units, Nonprofit Organizations</td>
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<td>Farmers Organizations, Community Organizations</td>
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<td>Jordan Valley Farmers Association, Ajloun Community Organization</td>
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level projects under development that improve water supplies by reducing leakage, or increase ground water pumping. In future work, the results can be further refined, however, we considered this set of large projects sufficient for the purpose of this analysis and to demonstrate the applicability of the method.

Traditionally, water sector projects have been developed on a least-cost basis [20]. Our discussion and surveys also confirmed the importance of this factor. We therefore, first plotted a cost versus supply curve for all combinations of the projects to analyze how it compared with current development (see Fig. 2). Of the five projects considered in our set, two have already been implemented: the Waste Water Treatment and Expansion (WWTE) project, and the Disi Aquifer Pipeline (DAP) that was recently completed and commissioned [21].

![Cost vs Water Supply for Combinations of Alternatives](image)

**Fig. 2. Least-Cost Development (Pareto Front) for Water Supply Projects.**

We find that the developed projects, WWTE (A5) and DAP (A4), lie on the pareto-curve (cost efficient frontier) of the decision space. Currently, the water supply system is in a state that consists of a combination of A5 and A4 (and is labeled as A5+A4 on the figure). In the future, the next possible states (in which three or four projects constitute the water supply system) can be E1 (A5+A4+A3), E3 (A5+A4+A2) or E5 (A5+A4+A3+A2) that lie on the pareto frontier.

If the curve is strictly followed in the future, then the new project (in case of E1) is A3 (the Agaba Desalination project) or (in case of E3) is A2 (the Jordan Red Sea Water Project). It should be noted that both of these projects are moving forward – and a major political agreement was recently made for the Jordan Red Sea Water Project [22]. It is interesting to note that the desalination projects, and especially the Red-Dead Canal project has been in planning documents for decades, however other options were developed since they were more cost effective. Since the lower cost options have been built, desalination projects are now being pursued.

Next, using data from Table IV, the matrix \( \mathbf{A} \) was compiled and the corresponding \( \mathbf{A} \) was computed using Eq. 1 - 3. We quantified the categorical variables with numbers 1, 3, and 9 for categories Low, Medium, and High.
A full-factorial combinatorial set of values of $w_i$ (of $w$ vector) was enumerated. For the set of five criteria ($m = 5$), and $dw = 0.05$, a total of 9113 valid values of $w$ (where Eq. 5 was satisfied) were used to evaluate the performance. In each case, $i^*$ (project with maximum performance) was determined, and for each project Eq. 7 was used to quantify likelihood of project selection. Fig. 3 shows the fraction of cases ($\pi_i$) for each projects when it was chosen as $i^*$ (i.e. had highest performance).

The results show that in a case of a choice between the five projects, WWTE had a probability of selection, $\pi_i$, of 83.1% (with $r = 7573$ and $q = 9113$), DAP had 9% ($r = 850$), and RDC had 7.9% ($r = 721$). This result compares well with actual decisions, where WWTE has been implemented.

The lower inset (with case of three projects, wherein WWTE and DAP are assumed to have been implemented), the highest likelihood is for JRSP from among the three alternatives of AD, RDC, and JRSP. Recent agreements between Jordan, Israel and Palestinian Authority show that plans for JRSP are advancing, and it seems highly likely to be implemented [22]. The completion of JRSP will add 120 million cubic meters of fresh water supply for Jordan.

The analysis shows (although with limited data) that the MCDA based results have good agreement with actual decisions. It is interesting to note, that if the pareto front of the cost-supply curve is strictly followed, the next project should have been the Aqaba Desalination plant (E1 in Fig. 2). In reality, the JRSP is advancing (E3 in Fig. 2), and the MCDA based assessment provides better indication of project selection.

In comparing recommendations of various published studies with actual decisions (as represented by commissioned projects), we find that optimal solutions – determined by sophisticated optimization techniques – may not be realized. For instance, in one case the recommendation of a recent study was to delay the Disi Pipeline [14], in reality the pipeline has been commissioned. This indicates a difference in local decision priorities that actually shape outcome versus a theoretical set of objectives.

For more comprehensive analysis, it is important to consider cases when decision preferences may shift. Given the difficulty estimating the probability vector $p$ of different scenarios related to population growth, increased migration, future reductions in water availability, etc., we constructed a $W$ matrix (see Eq. 8), with scenarios motivated by decision factor rankings provided by the different DMs. In the first scenario, the rank ordering of factors was supply, cost, political feasibility, and foreign investment potential and so on (see Table II). We constructed $W$ such that the values of $w_{ij}$ reflected the rank ordering of the factors. For instance, for the rank ordering given by DM1 (which we used to define scenario 1), the following condition was imposed:

$$w_{12} > w_{11} > w_{13} > w_{14} > w_{15} \quad (11)$$

where row index corresponds to scenario number, and column index corresponds to criteria number (shown in Table IV). In each scenario, Eq. 5 was also satisfied. Fig. 4 shows some sample cases for $w$. Note that there are many sets of values of $w$ that satisfy the criteria rank ordering (as shown in Eq. 11 for DM1).

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performance in the dark blue colored plots, and JRSP has highest expected performance in the light blue colored plots.

Fig. 5. Expected performance of projects under different scenarios of decision preferences.

In the first case where RDC is highest, inspection of Table II and Table IV shows that this is due to supply volume being the most important factor and this project provides the greatest supply volume as compared to other options. The expected performance of JRSP (a project that has recently undergone approval for development) is highest in fourth case (that corresponds to decision preferences provided by DM4 and DM5). The results clearly show that while case 4 matches best with current decisions, a shift in preferences can lead to different development decisions.

V. DISCUSSION AND FUTURE WORK

Water accessibility across different economic sectors of a region is defined by complex interactions of hydrological, economic, social, political and environmental factors. In this work we have focused on analyzing planned projects as a means of assessing mid-term (~5 to 10 years) availability provided through supply infrastructure development. We consider the question that given water supply possibilities (through a discrete set of projects), and knowing the decision criteria that are employed in a region, what projects are likely to be selected? The likelihood of selection of a project can be used to form an assessment of new supplies.

This approach offers a simple yet useful and traceable way for forming expectations. In arid regions, where there are few water supply options, the key real options are typically well known, and documented. It is unlikely that entirely new large-scale projects get conceptualized, garner political support and necessary funds, and get implemented in a short span of time. There is usually a good sense of feasible projects for the mid-term time horizon that analysts can use for applying this methodology.

While the simple approach offers utility, we also note its limitations. In our analysis we use a small sample survey – and recognize that it cannot provide statistically relevant data. However, in the case of public sector infrastructure projects, understanding the basis used by key decision makers can provide important insights. We also note that causality may not be interpreted from implemented projects – and implemented decisions cannot be the only basis for verifying underlying decision factors. This approach does not guarantee finding a unique set of criteria and their relative importance, however, we were able to provide some confirmation of the influence of decision factors as provided by decision makers on future projects.

Another key limitation is that of accurately populating the performance matrix. It is often difficult to find reliable data. Furthermore, conducting thorough stakeholder analysis is challenging and it may not always be possible to interview key decision makers.

We also note that it is possible for the criteria set to change or their relative importance to shift over time. Such changes can at times radically shift development pathways due to new and urgent priorities.

In future work, we will augment the analysis with a broader assessment of impact of variation in the criteria set and their importance on the outcomes. We will systematically consider the impact of sudden unforeseen shocks, crises, or disruptive changes that bring about important, large-scale shifts, through a more detailed analysis linking different scenarios and changes in the decision-making weights. Furthermore, we will expand the application of this method to other countries in the region (Oman and UAE) to further test and refine the methodology as well as to assess future water availability in those regions.

ACKNOWLEDGMENT

The authors thank Jade Salhab for assistance in organizing field interviews in Jordan. This work was supported by a research grant funded by BP in the Energy Technology Innovation Policy research group at John F. Kennedy School of Government, Harvard University. The findings, opinions, and conclusions expressed in this paper are those of the authors and do not necessarily reflect the views of the funding organization.

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