



39 **1. Introduction**

40 As global population and consumption of water rise, concerns that humankind is entering a new  
41 age of global water scarcity are increasingly widespread [*Liu et al.*, 2017]. To some, this rising  
42 water scarcity is worrisome because water is uniquely essential for myriad purposes – for  
43 drinking and critical domestic uses, as an input to food and industrial production processes, and  
44 for general human and ecological well-being [*Hanemann*, 2006]. Some predict that water’s  
45 essentialness will inevitably lead to a zero sum game and loss of livelihoods for specific users,  
46 widespread social destabilization, and environmental damage. Such warnings are perhaps most  
47 commonly heard in countries or regions where water scarcity is becoming a binding constraint to  
48 growth – the Middle East, Western United States, parts of Australia, and in river basins with  
49 intense water competition (e.g., the upstream Ganges, Nile, Mekong, or Tigris-Euphrates).  
50 Indeed, much of the globe already experiences acute economic water scarcity, due to a lack of  
51 high quality infrastructure and an inability of institutions to consistently provide the resource to  
52 end users [*Rijsberman*, 2006]. However, both institutional and infrastructure solutions, when  
53 designed and operated effectively, can dramatically improve water management, and thereby  
54 ease tensions.

55 At the same time, the effects of both water infrastructure and management interventions may  
56 vary, and need to be understood within their particular contexts. Learning which interventions  
57 work, and under what conditions, is vitally important, both for the very practical work of  
58 improving the performance of subsequent interventions within the specific context being  
59 targeted, and for applying broader lessons about drivers and impediments of key mechanisms of  
60 change to other contexts. Still, the methods to learn about impacts and mechanisms in the water  
61 sector – particularly the science of impact evaluation (IE) as applied to water and sanitation  
62 projects – remain imperfect. In short, the fundamental challenge facing researchers working on  
63 such IEs is to isolate causal effects of infrastructure from many other contemporaneous changes  
64 that affect water supply and sanitation services, water resources systems, and human welfare and  
65 well-being. In addition, as we discuss in this paper, the typical “gold standard” methodology for  
66 determining causal impacts, the randomized controlled trial (RCT) [*Bothwell et al.*, 2016; *Duflo*  
67 *et al.*, 2007], frequently faces validity, relevance, and practical challenges for the case of water  
68 infrastructure evaluation, and is arguably poorly suited to such applications. The considerations

69 we highlight echo and draw on various critiques of the pre-eminence ascribed to RCTs in IE  
70 literature more generally [*Deaton, 2009; 2019; Ravallion, 2018*].

71 This paper offers a perspective on the design and role of IEs for large water projects. We begin  
72 by discussing prior relevant literature, noting the growing body of rigorous work aimed at  
73 documenting the effects water and sanitation investments. The review reveals that prior  
74 evaluations typically provide answers to narrow questions and therefore face a challenge in  
75 providing a complete perspective on the effects of water infrastructure. This shortcoming is  
76 amplified as the scale of investment increases. In other words, the ability of IE to estimate  
77 impacts on well-being becomes more limited as the scale of the project under evaluation  
78 increases. This observation naturally leads to characterization and description of a number of key  
79 challenges that impede more holistic evaluation, including especially the tradeoffs or evaluation  
80 burdens inherent in such an investigation. Among these, we highlight the central IE challenge –  
81 the problem of rigorous causal attribution, but also focus on a set of other issues that deserve  
82 particular attention in the context of water infrastructure. Specifically, we emphasize the  
83 imperative of: engaging with project planners to conduct detailed *ex ante* mapping of an  
84 intervention’s theory of change; planning for spillovers and systems level changes, as well as the  
85 design contamination risks; considering distributional impacts (who wins and loses); monitoring  
86 non-monetary impacts including, e.g., quality of life changes; and finally, communicating clearly  
87 what aspects can and cannot be considered by a pragmatic and cost-effective IE approach.

88 To more effectively demonstrate these ideas, and shed additional light on the tradeoffs and  
89 particular problems that can emerge in such IEs, we discuss an effort to apply these principles to  
90 a US\$275 million infrastructure investment in urban areas of the Zarqa Governorate, Jordan. The  
91 locations targeted by this investment are lower-income areas in one of the most water scarce  
92 countries in the world. Through this application, we add to a surprisingly thin empirical literature  
93 on the *ex post* economic benefits of large water infrastructure [*Cox et al., 1971; Hanemann,*  
94 *2006*] and provide new evidence on the economic burden of unreliable water supplies in the  
95 Jordanian context. This evidence is timely because one of the Jordanian government’s major  
96 current objectives, supported by numerous policy reforms and changes in the water sector but  
97 few causal IEs, is to improve urban water security [*Royal Commission for Water, 2009*]. Various

98 reforms are occurring in a complicated political economy context that constrains the use of price  
99 instruments, owing to widespread popular opposition to higher water bills [*Klassert et al.*, 2018].

100 The remainder of the paper is organized as follows. In Section 2, we discuss prior literature on  
101 the impacts of large infrastructure, with a particular focus on the water and sanitation sector and  
102 especially urban piped water and sewer improvement. Section 3 describes the general challenges  
103 that confront evaluators of such projects, and Section 4 presents the example urban water  
104 infrastructure application that clarifies the nature of many of these problems. Section 5 presents  
105 an integrated view of the results from that application, and Section 6 discusses these findings and  
106 offers general reflections on the value of IE methods for assessing the impacts of water  
107 infrastructure.

108

## 109 **2. Background: Prior literature on the impacts of infrastructure projects**

110 We begin this section by summarizing prior reviews and perspectives on IEs of large  
111 infrastructure, before turning more specifically to those in the water and sanitation sector. We  
112 draw on several notable reviews or surveys of infrastructure evaluations in developing countries,  
113 which is our contextual focus given the important potential link between such investments and  
114 economic development outcomes [*Brakarz and Jaitman*, 2013; *Estache*, 2010; *Raitzer et al.*,  
115 2019; *Sawada*, 2015]. *Estache* [2010] highlights the relative paucity of large infrastructure IEs  
116 (in water, transport, and electricity) relative to those of interventions in other sectors, especially  
117 health and education, despite a high policy demand for rigorous evidence. He largely attributes  
118 this dearth of evidence to methodological challenges that push academic researchers to forgo  
119 evaluation efforts: problems of non-random assignment (which threaten causal inference); the  
120 fact that their benefits take a long time to materialize; and the challenge of dealing with complex  
121 feedbacks linking economic development and infrastructure. The review work also highlights  
122 that impacts are highly variable across intervention types, technology, social and institutional  
123 contexts, and the information or knowledge that target beneficiaries possess, and emphasizes that  
124 good infrastructure IEs must explicitly address spillovers (i.e., effects on populations not  
125 considered to be directly “treated” by the interventions).

126 These challenges notwithstanding, existing work focusing on road construction and extension of  
127 the electrical grid in developing country contexts emphasizes the importance of such  
128 infrastructure to economic growth. The evidence is perhaps strongest for roads and connectivity  
129 [Aggarwal, 2018; Dercon *et al.*, 2009; Ghani *et al.*, 2016; Jedwab and Storeygard, 2022], though  
130 several studies note substantial heterogeneity in outcomes. Analyzing the effects of road  
131 construction in 39 African countries, Jedwab and Storeygard [2022], for example, find that  
132 connected cities grow faster than unconnected ones, but that the elasticity of this growth is  
133 greater for smaller and more remote cities, and weaker in politically favored or agriculturally  
134 suitable areas. The political economy result is particularly important in light of the preferential  
135 siting of infrastructure in favored locations [Blimpo *et al.*, 2013]. Growth also appears to be  
136 primarily driven by rural to urban migration. In rural settings, however, the complementary  
137 conditions needed for positive impacts on village economies (mainly income and wealth gains)  
138 may remain elusive even when market connectivity is enhanced [Asher and Novosad, 2020; Mu  
139 and Van de Walle, 2011]. Other work highlights the importance not just of the connectivity  
140 provided by roads, but in their quality [Casaburi *et al.*, 2013]. Evidence of the positive benefits  
141 of rural electrification, by contrast, is somewhat more mixed. While several studies in middle-  
142 income countries have shown positive impacts for some types of outcomes [Dinkelman, 2011;  
143 Lipscomb *et al.*, 2013], reviews synthesizing the broader literature suggests rather muted  
144 impacts, especially in comparison to investment costs, and again points to the importance of  
145 heterogeneity [Bos *et al.*, 2018; Jeuland *et al.*, 2021b; Lee *et al.*, 2020; Peters and Sievert, 2016].

146 Turning to water infrastructure specifically, despite the obvious importance of such investments  
147 to confront economic water scarcity [Molden, 2013], there are surprisingly few evaluation  
148 studies outside of rural settings. Indeed, most related literature analyzes the impacts of non-  
149 network water, sanitation and hygiene (WASH) solutions, as commonly implemented in rural  
150 areas. Moreover, even in the water sector, many infrastructure IEs focus on child health and  
151 school attendance as the main primary outcomes, with time savings, social/gender inclusion and  
152 political participation, privacy, and various other welfare indicators garnering less attention  
153 (income, consumption, coping costs) [Estache, 2010]. Common methods deployed in such  
154 studies include rigorous field-based IE designs [Duflo *et al.*, 2015; Hammer and Spears, 2016;  
155 Lokshin and Yemtsov, 2003; Meeks, 2017; Pattanayak *et al.*, 2010], regression methods applied  
156 to cross-sectional or panel data [Esrey, 1996; Pickering and Davis, 2012; Whittington *et al.*,

157 1990], and reviews and meta-analyses that synthesize evidence. The latter are almost exclusively  
158 focused on health [*Waddington et al.*, 2009; *Wolf et al.*, 2022]. Still in rural areas, a minority of  
159 studies examine the effects of piped water [*Brown et al.*, 2013], where the comparison group is  
160 typically comprised of households without access to network services. Finally, much of this  
161 work speaks to distributional effects in their emphasis on particular sub-populations of  
162 beneficiaries, especially children (for health improvements) (e.g., Hammer and Spears [2016]),  
163 and women (concerning time savings) (e.g., Pickering and Davis [2012]; Meeks [2017]).

164 For investments in piped water and sanitation, existing evidence primarily comes from urban or  
165 peri-urban settings, and mostly covers the extension of services to unconnected populations, and  
166 benefits produced over the near term. Galiani et al. [2009] found that increased water access in  
167 shantytowns can lower household coping costs, leading to increased savings of money and time;  
168 a set of somewhat different studies examine the impacts of general slum upgrading that included  
169 piped water connections among other improvements [*Soares and Soares*, 2005]. Positive but  
170 non-statistically significant effects on income and labor allocation were also found in a  
171 difference-in-differences analysis of secondary data covering both urban and rural Vietnam  
172 [*Nguyen Viet and Vu*, 2013]. Time savings – along with reduced intra-household conflict over  
173 water – were also an important outcome of experimentally-induced increases in piped water  
174 connections in urban Morocco [*Devoto et al.*, 2011].

175 Health gains have also been identified in several settings and over different time periods by  
176 examining the introduction of piped water and sanitation service access and various measures of  
177 illness and mortality over time [*Alsan and Goldin*, 2015; *Galiani et al.*, 2005; *Gamper-*  
178 *Rabindran et al.*, 2008; *Jalan and Ravallion*, 2003]. Health improvements were not detected in  
179 the aforementioned study from urban Morocco, however, perhaps because transmission of  
180 diarrheal disease in target communities was low to begin with, or because of network water  
181 quality problems [*Devoto et al.*, 2011]. To be sure, beneficiaries often worry about the quality of  
182 piped water: respondents to a survey in Nigeria were willing to pay a large premium for water  
183 from vendors, despite major structural improvements of the water system [*Whittington et al.*,  
184 1991].

185 Efforts to improve existing piped systems' quality or technology, rather than new connections,  
186 have been less frequently studied. In this domain, the phasing in of better water treatment –

187 specifically chlorination – was found to sharply reduce mortality in the US in the early 20<sup>th</sup>  
188 Century [*Cutler and Miller, 2004*]. Also related to quality and reliability, however, is whether  
189 improvements can be sustained over time while keeping costs manageable given the  
190 development context of a particular place. Here the experience of urban Yemen is instructive;  
191 researchers found that access to piped water supply actually *worsened* health outcomes, and  
192 attributed this to rationing and a buildup of pollution in the network [*Klasen et al., 2012*]. Such  
193 issues may be a particular challenge in developing countries where network water is rationed,  
194 and where low water tariffs have been linked to shortages and reduced long-term utility  
195 performance [*Bucknall et al., 2007; Foster and Briceño-Garmendia, 2009; Jeuland, 2012*]. This  
196 may lead to perpetuating an infrastructure quality trap [*Burt et al., 2018; Ercumen et al., 2015;*  
197 *McRae, 2015*].

198 Compared to piped water, there is much less evidence related to urban sewerage, and most of  
199 that existing evidence focuses exclusively on health. Two key references are Waddington et al.’s  
200 [2009] and Wolf et al.’s [2022] systematic reviews. These studies identified six IE estimates in  
201 total in this category of interventions, of which five found no significant impacts on health  
202 [*Galdo and Briceño, 2011; Klasen et al., 2012; Kolahi et al., 2009; Pradhan and Rawlings,*  
203 *2002*], versus the one that did [*Moraes et al., 2003*], though pooled estimates were somewhat  
204 more positive [*Wolf et al., 2022*]. Historical studies from OECD countries are more definitive in  
205 relating diffusion of sewers to declining mortality, but are also somewhat limited in number  
206 [*Alsan and Goldin, 2019; Kesztenbaum and Rosenthal, 2017*].

207 Importantly, all of the micro evidence on the impacts of water and sanitation infrastructure  
208 discussed above concerns impacts on households, though older literature from developing Asia  
209 and OECD countries has investigated linkages between water infrastructure and national or  
210 regional economic income [*Cicchetti et al., 1975; Uchimura and Gao, 1993*]. Additional  
211 descriptive (but not counterfactual-based) evidence related to the impacts of piped water and  
212 sewer networks also supports the idea that urban water supply investments may reduce firms’  
213 input costs [*Schwartz and Johnson, 1992*]. The idea here is that beneficiary firms will potentially  
214 respond to reductions in costs with expanded production and employment, investment, and  
215 profit, or by reducing output prices to the benefit of consumers. These linkages are perhaps  
216 especially important where scale economies for piped services are large, water is a major input to

217 production, and current alternative sources are inadequate [*Schwartz and Johnson, 1992*]. Like  
218 households, however, demand for piped water supply improvements may be limited by concerns  
219 over quality [*Davis et al., 2001*].

220 All in all, IE literature on the effects of regional or urban water and sanitation infrastructures  
221 remains thin. Considering the relative richness of evaluation studies of stand-alone systems in  
222 rural areas, this lack of evidence cannot reasonably be attributed to the lack of importance of  
223 such evidence. Rather, it seems much more likely that rigorous counterfactual studies of urban  
224 and network improvements are difficult to implement. At least as importantly, this literature is  
225 highly fragmented: researchers in the studies reviewed above typically look at very specific or  
226 narrow sets of outcomes. This is despite the fact that network water sector interventions logically  
227 affect a range of outcomes, as documented above, that range from household health and well-  
228 being to productivity and net income improvements, and extend to beneficiary businesses and  
229 even the utilities providing such services.

230 We thus conclude this review with a brief summary of important challenges facing water  
231 infrastructure IEs. First, due to the nature of the scale of such interventions, which often consist  
232 of large, multi-pronged and overlapping activities, impacts can be difficult to attribute to specific  
233 investments. Relatedly, observed differences in outcomes for beneficiaries may be the stem from  
234 a combination of factors occurring through multiple channels. Some changes may be unrelated to  
235 the intervention itself, but may nonetheless mediate its outcomes in ways that are crucial to  
236 understand for sound policy-making. Second, the potential outcomes of urban water  
237 interventions are many and diverse both in type and in magnitude, which creates practical  
238 challenges related to measurement and the statistical power of the evaluation. Third, urban water  
239 interventions may cover all residents in an area (making it difficult to find a suitable comparison  
240 group), or alternatively target populations and locations that are very different from those who  
241 are untargeted [*Lokshin and Yemtsov, 2003*]. Fourth, infrastructure development alone may not  
242 deliver quality and reliability, particularly in the long term, if systems are poorly managed and  
243 operated [*Zérah, 1998*]. We reflect more critically on these challenges and their implications for  
244 IEs in the subsequent sections.

245

246 **3. The central challenge facing water infrastructure IEs, and alternative methods for**  
247 **addressing it**

248 To discuss the challenges identified above more formally, this section begins with a description  
249 of the central problem for researchers working to evaluate water and sanitation infrastructure (or  
250 really any development) interventions. We then provide an overview of the most commonly  
251 implemented approaches, offering reflections on their relative strengths and weaknesses. We  
252 close the section with a call for more mixed-methods IE designs that reveal a more complete set  
253 of consequences from such projects, before presenting a real-world application of that idea.

254 The basic problem facing any causal IE is to estimate the difference between what happened as a  
255 result of an intervention with what would have occurred in its absence. We present a simple  
256 framework to illustrate. Consider the outcome of interest  $Y$  and a dichotomous indicator for  
257 exposure to an infrastructure intervention  $I$ , which takes a value of 1 if the intervention occurs,  
258 and is zero otherwise. The level of the outcome given “treatment” (or  $I = 1$ ) is defined as  $Y^1$ , and  
259  $Y^0$  if  $I = 0$ . The impact caused by the infrastructure is then just the average treatment effect on the  
260 treated (ATT):

261 
$$ATT = E(Y^1 - Y^0 | F = 1). \quad (1)$$

262 In this formulation, the major challenge is finding an unbiased method for approximating the  
263 counterfactual outcome  $Y^0$  had the unit not been treated, which by definition cannot be observed  
264 directly. In a naïve design that simply compares observations that are targeted to receive the  
265 improvement against observations that do not, we observe:

266 
$$E(Y^1 | F = 1) \text{ and } E(Y^0 | F = 0). \quad (2)$$

267 Taking the difference between these terms we observe that:

268 
$$E(Y^1 | F = 1) - E(Y^0 | F = 0) = [E(Y^1 | F = 1) - E(Y^0 | F = 1)] + [E(Y^0 | F = 1) -$$
  
269 
$$E(Y^0 | F = 0)] = E(Y^1 - Y^0 | F = 1) + [E(Y^0 | F = 1) - E(Y^0 | F = 0)]. \quad (3)$$

270 This approximation based on treated and non-treated groups deviates from the ATT from  
271 equation (1) by the two final terms in equation 3, which represent the differences in outcomes  
272 across the comparison units in the absence of the treatment, and which helps clarify the  
273 evaluator’s problem with selection bias. In nearly all cases, to achieve high impact, water and

274 sanitation infrastructure is designed and delivered precisely to the groups that are hypothesized  
275 to benefit the most from such projects; this will lead to upward bias in the treated-untreated  
276 estimate of the ATT because  $E(Y^0|F = 1) > E(Y^0|F = 0)$ . On the other hand, if the  
277 infrastructure is targeted at groups that somehow benefit less from such investments (perhaps  
278 due to their lower income or access to other resources needed for development), the ATT  
279 estimate using Equation 3 may be biased downwards. Thus, one of the major challenges plaguing  
280 observational studies of the impacts of improved water and sanitation services on outcomes is  
281 that beneficiaries who receive improved services tend to be systematically different from those  
282 who do not (in terms of socio-economic status (SES), risk-altering behaviors, unobserved  
283 preferences for health, or myriad other ways), rendering comparisons of those with and without  
284 access suspect.

285 Rigorous IE methods are meant to minimize this risk of bias, by constructing more comparable  
286 groups of treated and untreated observations. In the simplest, most straightforward experimental  
287 case, we can assert that  $[E(Y^0|F = 1) = E(Y^0|F = 0)]$ , at least in a statistical sense. But  
288 random assignment is rarely feasible or desirable for large infrastructures, so a variety of quasi-  
289 experimental methods are more commonly utilized (Table 1). The idea is to leverage situations  
290 that lead to plausibly exogenous (“as if randomized”) variation in exposure to infrastructure, or  
291 to apply statistical methods to isolate impacts by creating more comparable groups. A thorough  
292 review of how various approaches mitigate selection bias is beyond the scope of this article, but  
293 prominent references describing each are provided as a guide in Table 1. There we also specify  
294 and comment on three criteria, discussed also in what follows, that help clarify the  
295 appropriateness of these methods; comments pertinent to water infrastructure IEs are especially  
296 emphasized.

297 The first such criterion pertains to the validity of the design for proper causal inference, which  
298 refers to “internal validity”, or the degree of confidence that measured effects are in fact due to  
299 the investment in question. Besides the obvious issue of selective targeting discussed above,  
300 infrastructure projects such as network water and sanitation give rise to several other common IE  
301 challenges. For one, even the most rigorous experimental or quasi-experimental designs may  
302 suffer from confounding by (unobserved) differences that are unknown to the evaluator. Such  
303 differences can arise due to unbalanced randomization (especially in small sample RCTs) or

304 from a lack of full accounting for such differences (in quasi-experiments), when they correlate  
305 with targeting criteria. In difference in differences (DiD) designs, the major concerns are  
306 unobserved time-varying unobservables, which can give rise to non-parallel trends in the  
307 treatment and comparison groups. In matching designs, the conditional independence assumption  
308 requires that the variables that affect treatment assignment and treatment-specific outcomes are  
309 fully observable, such that any dependence between them is removed by conditioning on these  
310 factors [*Rosenbaum and Rubin, 1985*].

311 As complex and large interventions, infrastructure projects may also generate substantial  
312 spillovers, whereby untreated beneficiaries, who provide the counterfactual in an IE, are  
313 indirectly affected, typically as a result of behavioral responses. For example, a piped water  
314 improvement may induce some people to move – due to changes in asset values or individuals’  
315 desire to capture its benefits. This may reduce resource pressure in comparison areas, or lead  
316 utility personnel to adjust system operations in a way that affects those in unimproved areas.  
317 Spillovers can also play out through general equilibrium effects or distributional channels, for  
318 example, if informal water and sanitation service providers like water tankers respond to  
319 improvements and reduced service demand by lowering their prices, thereby generating benefits  
320 for untreated (comparison) consumers. An additional issue is selective attrition, whereby  
321 beneficiaries observed after an intervention may be more likely to provide data than those  
322 unserved (perhaps because beneficiaries are more willing to complete follow-up surveys than  
323 those in the comparison group). This is a problem when those providing data are not a random  
324 subsample of the larger population they were intended to represent, if, for example, only the  
325 most cooperative people participate.

326 The second criterion in Table 1 refers to the relevance of IE evidence that is produced. Here,  
327 evaluations generally place the greatest emphasis on “external validity”, or on the  
328 generalizability of results to other groups and settings. Besides generalizability, it is important to  
329 emphasize relevance to answering decision makers’ most crucial questions, which may be  
330 constrained by a particular method or set of methods. Related to these issues, Table 1 discusses:  
331 a) whether the IE pertains to an artefactual (largely researcher-constructed) or real-world (as  
332 implemented) situation; b) the population to which the evaluation evidence applies (i.e., whether  
333 it is all of those treated or some unique sub-sample); c) the likelihood that the produced evidence

334 will be convincing to domain and IE experts; d) whether results are highly conditioned by  
335 assumptions that cannot be easily substantiated; and e) whether the evidence is likely to be  
336 precise from a statistical perspective. The answers to these questions are often related, with  
337 tradeoffs among them.

338 More specifically, real-world infrastructure interventions seldom give rise to situations that can  
339 be assessed with the methods deemed most convincing by IE experts (item *a* in the previous list).  
340 Conducting an RCT of piped water extension efforts, for example, is typically infeasible: It  
341 requires unprecedented coordination of costly investments across a large range of eligible  
342 locations, from which a treatment group is randomly selected. In addition, implementers often  
343 view randomization as arbitrary and poorly targeted, even in situations that give rise to  
344 opportunities for such a design. On the other hand, natural experiments, instrumental variables,  
345 and *ex post* regression are commonly applied to fully real world situations. The choice among  
346 these approaches demands careful thinking about the confounding and selection threats that face  
347 real world infrastructure evaluations. The corollary is that causal evidence derived from the latter  
348 methods is more often in doubt, relative to the “gold standard” RCT (issue *c*), whose results  
349 within a given experimental population are not conditioned by assumptions embedded within the  
350 deployed analytical methods (issue *d*). Nonetheless, critiques of RCTs often highlight the  
351 difficulties that arise in transferring their results across contexts, given the (usually) limited  
352 attention paid to contextual or institutional assumptions embedded in an experiment [*Peters et*  
353 *al.*, 2018]. The reality is that the vast majority of RCT evaluations are researcher-driven and  
354 artefactual, typically designed to test a narrow set of causal relationships.

355 The issues of the population for which the treatment effect is measured (issue *b*) and statistical  
356 precision (issue *e*) also have considerable importance. With the former, it is important to  
357 distinguish between estimates of intention to treat (ITT) impacts, which are representative of the  
358 entire targeted population and account for partial compliance or uptake, and treatment on the  
359 treated (ToT) impacts, which pertain only to those who take up the intervention, or to a subset of  
360 the treated. ToT impacts have more limited relevance when there is strong selection into  
361 treatment, or when the selection processes are unclear (e.g., as when instrumental variables  
362 methods are used) and give rise to a very specific local average treatment effect (LATE). ITT  
363 impacts, on the other hand, will be less transferrable if the processes that determine compliance

364 and uptake rates do not generalize across settings. At the risk of oversimplifying an extensive  
365 discussion in the literature, we can assert that some methods produce more population-  
366 representative estimates (e.g., DiD, RCT, natural experiment), while others allow only treatment  
367 effects measurement in a very specific sub-population (e.g., instrumental variables (IV),  
368 regression discontinuity (RD)), with others (e.g., matching) falling somewhere in between.

369 Finally, the question of statistical precision for measuring impacts depends primarily on  
370 appropriate sampling rather than on analytical methods. But the truth is that water and sanitation  
371 infrastructure has the potential to influence a host of indicators through various channels. As  
372 such, relying on traditional single-method IEs will typically fail to capture important phenomena.  
373 For example, water and sewer interventions may benefit households in myriad ways (reduced  
374 coping costs, increased water consumption, improved health, greater productivity, etc.) when  
375 water scarcity is relieved [*Waddington et al.*, 2009; *Zwane and Kremer*, 2007], or benefits may  
376 flow mainly to businesses using water as an input to production [*Schwartz and Johnson*, 1992] or  
377 to a utility, if the latter collects more revenue from water sales [*Jeuland et al.*, 2020a]. Tracking  
378 impacts on these various groups may require very different data collection efforts, or even  
379 different methods, and may require combining natural and artefactual experiments [*Sawada*,  
380 2015]. Moreover, many network interventions (such as piped water and sewer investments) have  
381 multiple components, and may therefore require complex designs that combine several research  
382 strategies.

383 The third criterion in Table 1 refers to practical and logistical considerations that emerge from  
384 the implementation of specific IE methods. The issues covered include a) costs; b) risks and  
385 adaptability to mitigate contamination arising from the spread of the treatment to comparison  
386 areas; c) need for coordination with project planners; d) interpretability and transparency of  
387 results; e) data needs from the pre-project period; and f) applicability or flexibility for covering  
388 complex interventions or theories of change. In general, the most rigorous methods – those that  
389 maximize internal validity – tend to be more costly, require greater coordination with planners,  
390 have greater pre-project data requirements (except for the RCT), and face the most severe  
391 contamination threats. This helps to explain why many studies deploying such methods are  
392 artefactual, researcher-driven ones, and why they are so rare in infrastructure evaluations.  
393 Moreover, there is frequently a tradeoff between transparency and flexibility or adaptability. This

394 is demonstrated by comparing flexible IV, modeling and matching methods, which rely more  
395 heavily on analysts' judgement, to less flexible RCT, RD, and natural experiment approaches.

396

#### 397 **4. Design application: The Millennium Challenge Corporation Jordan Compact**

398

399 With this general discussion in mind, we next turn to an evaluation application that illustrates the  
400 tensions and choices more concretely. This example consists of an integrated and holistic  
401 infrastructure improvement for urban water and sewer services in Zarqa, Jordan, a populous area  
402 in one of the most water-poor countries in the world [*Haddadin, 2006; Schyns et al., 2015*].

##### 403 *4.1. Context*

404 In 2017, the year in which the investment we describe below was fully online, Jordan had per  
405 capita annual renewable water resources of 96 m<sup>3</sup>, nearly 3.5 times below the average of  
406 other arid Middle Eastern countries, and more than 50 times lower than the world average [*FAO,*  
407 *2019*]. Recent population growth, urbanization, an influx of refugees from Iraq and Syria, and  
408 extremely high water losses, have intensified this strain [*Hashemite Kingdom of Jordan, 2016*].  
409 These growth trends and water losses have been particularly high in urban Zarqa, the region  
410 targeted by the infrastructure investment we consider. For example, non-revenue water (NRW)  
411 has been estimated to exceed 50% over the recent period, well above global and national  
412 averages [*Jeuland et al., 2020b*].

413 Most of the population in Zarqa Governorate lives in Zarqa City (the second largest city in  
414 Jordan; population ~802,000) and Ruseifa (4th largest city; pop ~482,000), both of which lie in  
415 the Zarqa River Basin. Inhabitants of these cities have considerably lower income than those in  
416 neighboring Amman. Water supply is highly rationed; both households and businesses  
417 experience burdensome and routine water shortages and received water for only about 24 hours a  
418 week prior to the infrastructure improvements, in 2015 [*Orgill-Meyer et al., 2018*]. In addition,  
419 though more than 99% of households have access to piped water, only about 70% had sewer  
420 connections prior to the new investments, and nearly 30% were thought to consume less than the  
421 minimum amount of water that the World Health Organization considers necessary (60L/capita-

422 day) for personal hygiene and food safety [MCC, 2009].<sup>1</sup> In addition to water scarcity and  
423 reliability problems, households perceive the quality of utility water in Zarqa to be poor. Small  
424 businesses have lower reliance on the piped water (48%) and sewer (64%) networks than  
425 households, reportedly due to high connection fees but also unwillingness to pay tariffs for water  
426 consumption that cross-subsidize domestic users [Jeuland et al., 2015]. Though Jordanians pay  
427 relatively high water tariffs compared to populations in neighboring countries, the utility in  
428 Zarqa (now called Miyahuna-Zarqa) does not fully cover its costs, and there has been long-  
429 standing resistance to water tariff increases [Pitman, 2004; Sommaripa, 2011].

#### 430 4.2. *The intervention: The Jordan Compact*

431 The Millennium Challenge Corporation – Millennium Challenge Account Jordan Compact  
432 (MCC-MCA-J JC, hereafter referred to as the JC) was developed to address Zarqa’s most  
433 important water and sewer network deficiencies.<sup>2</sup> MCC largely funded the water investments and  
434 worked with the Government of Jordan (GoJ) throughout a detailed project identification and  
435 preparation period.<sup>3</sup> Construction began in 2014, implementation was largely successful, and the  
436 infrastructure handover had been completed as planned by the end of 2016. The final JC included  
437 three inter-linked projects:

438 (i) The Water Network Project (WNP) comprised two activities. The first was rehabilitation and  
439 restructuring of water supply transmission and distribution infrastructure in Zarqa and Ruseifa,  
440 and replacement of domestic water meters. The aim was to improve overall water efficiency  
441 through reduction of physical water losses and a transition from periodic distribution under high  
442 pressure to more consistent, gravity-fed distribution. The second activity, Water Smart Homes  
443 (WSH), aimed to improve household water storage and sanitation through a general outreach  
444 campaign, and to deliver infrastructure and technical assistance to the poor.

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<sup>1</sup> Those lacking sewer connections usually rely on septic tanks that regularly require evacuation by tanker trucks, and also raise the risk of network contamination through infiltration of depressurized pipes.

<sup>2</sup> MCC is a U.S. foreign assistance agency that aims to fight global poverty by focusing on good policies, country ownership, and results. MCC provides time-limited grants to recipient countries, and administers the MCA. Thus, when a country is awarded a compact, it sets up a local MCA entity to manage and oversee all aspects of project implementation. Monitoring of funds is rigorous and transparent, often through independent fiscal agents.

<sup>3</sup> This preparation phase involved an analysis that was aimed at identifying key challenges to economic development in Jordan (a constraints analysis), identifying a technical solution to some of those challenges, and conducting feasibility and economic analyses of its anticipated impacts. More information on how MCC works can be found at <https://www.mcc.gov/>.

445 (ii) The Wastewater Network Project (WWNP) provided the expansion, rehabilitation and  
446 reinforcement of the wastewater network in Zarqa Governorate, and aimed to increase the  
447 capture of municipal wastewater and improve wastewater system efficiency.

448 (iii) The As-Samra Expansion Project (AEP) was designed to raise the capacity of the existing  
449 wastewater treatment plant serving this region, to allow treatment of additional wastewater  
450 volumes resulting from population growth in Amman and Zarqa, and from the aforementioned  
451 WNP and WWNP investments.

452 The stated goals of the JC – to reduce poverty and stimulate economic growth in Zarqa – were  
453 ambitious. These goals were to be achieved by increasing urban water supply, because water  
454 scarcity was deemed a key constraint inhibiting the area’s development. *Ex ante*, planners  
455 believed that the investment would result in benefits through three main channels: two water  
456 substitution mechanisms, plus household cost savings from reduced dependence on septic tanks  
457 [Jeuland *et al.*, 2020b]. The first (primary) water substitution effect relied on capture and reuse  
458 of the additional volumes of treated wastewater resulting from greater water consumption and  
459 sewage collection in Zarqa, for irrigation in the Jordan Valley (JV), thereby facilitating  
460 reallocation of scarce freshwater sources for high value urban uses, while maintaining high value  
461 agriculture. Figure 1 provides a visual and qualitative representation of this primary substitution  
462 mechanism. The second (secondary) substitution effect posited that households in Zarqa would  
463 switch away from high-cost non-network water vendors (i.e., distribution shops and tankers) to  
464 greater use of cheaper network water, once urban water supply became more reliable.

#### 465 4.3. *Development of the IE design and rationale for the final approach*

466 While these three main benefit mechanisms of the JC are simple and intuitive, complex  
467 infrastructure projects can entail a slew of varying short and long-term effects that demand more  
468 careful examination. To consider the possibilities, the evaluation work began with consultations  
469 with a wide range of key stakeholders – representing different GoJ agencies, the implementing  
470 unit at the MCA-J, and the MCC. This engagement revealed that different parties had somewhat  
471 divergent perspectives on the project’s most critical aspects, and that many potential sub-  
472 mechanisms and assumptions underpinned or supplemented the three main channels described  
473 above. For most real world IEs, participatory elicitation of a program theory of change and the  
474 context in which it operates, as conducted for this case, is of preminent importance prior to

475 design and data collection [White, 2011]. A literature review of prior related international  
476 experiences helped to identify additional potential channels of impact and clarify planners'  
477 assumptions, and resulted in a more comprehensive project logic (Figure 2).<sup>4</sup>

478 After development of this more complete project logic, the next step was to develop an IE that  
479 was practical, rigorous, and relevant to assessing the most important aspects of the investment.  
480 The two primary considerations in this design work were to specify the appropriate scope (e.g.,  
481 populations and locations requiring study) and the nature of approximation of the non-  
482 intervention counterfactual. Given the integrated nature of the program's economic logic and its  
483 effects on multiple sectors as discussed above, the design endeavored to incorporate a diverse set  
484 of affected parties and geographies (Table 2 summarizes these populations; Figure 3 shows the  
485 geographic scope of data collection, Figure 4 clarifies the timeline of data collection events  
486 relative to infrastructure construction, and the supplementary materials provide additional  
487 details). Challenges pertaining to generalizability, statistical power, proper accounting for  
488 spillovers, flexibility to tackle changes in program implementation, and adequate  
489 contextualization and control for non-intervention confounders all helped to inform the final mix  
490 of evaluation components, methods, and data or measurement types. These choices were also  
491 subject to limitations on the overall evaluation budget.

492 In brief, the balancing of these aspects ultimately led to a design with three core components that  
493 combined several of the methods discussed in Table 1: i) DiD assessment of the infrastructure  
494 improvements' effects on urban populations (households and small enterprises, accounting for  
495 expected declines in meter accuracy that occur where service is intermittent) in urban Zarqa  
496 Governorate, with matching to ensure comparability of those with varying exposure to the JC; ii)  
497 counterfactual water balance modeling and DiD analysis of the infrastructure investments'  
498 effects on farmers (stemming from hypothesized increases in treated effluent inflows to the

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<sup>4</sup> It is worth highlighting several additional channels and key assumptions that the evaluation team deemed to be potentially consequential at the time of design (Jeuland et al. 2020). First, some hypothesized that households might experience changes in health and general well-being arising from shifts in sourcing and in consumption levels or other behaviors targeted in the outreach campaigns organized around the JC. Second, it was deemed possible that the water utility would capture many of the benefits of the investment, with direct implications for cost recovery and utility performance, and indirect implications for public debt and longer-term quality of service delivered to households. Third, there was a sense that firm-level water decisions might change, and perhaps especially those of water shops and tankers, with effects on the distribution of economic outcomes. Finally, the switch in irrigation water sourcing was posited to have implications for farmers growing salinity-sensitive crops (e.g., citrus) in the Jordan Valley.

499 Jordan Valley); and iii) tracking of Zarqa utility performance over time, relative to that of two  
500 other similarly corporatized urban utilities in Jordan, in Amman and Aqaba. Additional details  
501 on the sampling and comparability, as well as approximate locations, of the IE's counterfactual  
502 populations are provided in the supplementary materials to this article (in Appendix 1). Needless  
503 to say, the use of such diverse combined methods scarcely appears in infrastructure evaluations  
504 present in the literature, and represents an important contribution of this application.

505 Several of the data collection activities shown in Table 2 were not central to the evaluation  
506 design but were included to deepen insights emerging from these 3 core IE components.  
507 Specifically, an endline cross-sectional survey of water vendors was included to both confirm  
508 and clarify distributional impacts related to the secondary substitution mechanism described in  
509 Section 4.2 (i.e., the posited shift away from expensive vendor water). Also, owing to the  
510 security crisis in neighboring Syria, Jordan received a major inflow of refugees that was  
511 contemporaneous with JC implementation; we carried out a refugee survey to understand the  
512 additional demand pressure this entailed, whether it was felt evenly across locations with  
513 differential exposure to the JC improvements, and to control for those differences, to the extent  
514 that they were important. Finally, we employed qualitative key informant interviews to better  
515 understand implementation fidelity and contextualize the main quantitative results.

#### 516 *4.4. Econometric and modeling analysis of impacts*

517 The DiD analyses for the three affected populations identified above – households in Zarqa,  
518 small enterprises in Zarqa (equation 4), and irrigators (equation 5) located downstream of the As  
519 Samra wastewater treatment plant – follow similar econometric specifications, which leverage  
520 panel data and net out time-invariant unobserved differences across groups that could otherwise  
521 affect or confound interpretation of the impacts of the JC:

$$522 \quad Y_{ijt} = \alpha + \gamma_1 T_t + \delta d_j + \kappa_t T_t \cdot d_j + \beta X_{ijt} + v_i + \delta_{ijt}, \quad (4)$$

523 where  $Y_{ijt}$  is the outcome of interest for household/farm/enterprise  $i$  in zone  $j$  at time  $t$  (with  $t = 0$   
524 before intervention, and 1 after intervention);  $d$  is a vector of dummy variables that are equal to 1  
525 if household/farm/enterprise  $i$  is in treatment area  $j$  and 0 otherwise;  $T_t$  is a vector of dummy  
526 variables that are equal to 1 for the period in which the data was collected and 0 otherwise,  $X_{ijt}$  is  
527 a vector of time-varying control variables that may affect the outcome for unit  $i$  in zone  $j$  at time

528  $t$ ;  $\nu_i$  is a fixed effect for household/farm/enterprise  $i$ ; and  $\delta_{ijt}$  is a time-varying error term. The  
529 coefficient  $\kappa_j$  measures the “treatment effect” or the change in outcome  $Y$  for  
530 households/farms/enterprises in group  $j$  relative to that in the omitted comparison group. This  
531 estimate is unbiased so long as the error term  $\delta_{ijt}$  is uncorrelated with treatment status. To test  
532 robustness, we estimated models with and without individual unit fixed effects, and with and  
533 without time-varying controls  $X_{ijt}$ , though our preferred specifications include both of these.  
534 Standard errors were clustered at the level of the sampling cluster, generally the Census block.

535 The “treatment effect” measured among households and enterprises is derived from distinct sub-  
536 samples – for those exposed to different JC interventions and combinations – of treated and  
537 comparison observations that were randomly sampled from the zones in these two categories,  
538 selected for comparability using matching prior to implementation and baseline data collection.  
539 Specifically, zones specified as receiving each of the WNP only, WWNP only, and both WNP  
540 and WWNP improvements, were matched to separate samples of comparison zones using  
541 Census data and a 1-1 nearest-neighbor Propensity Score Matching (PSM) approach (see  
542 Appendix 1 for additional details on this sample construction). Thus, the effects of these 3  
543 different intervention combinations are estimated from 3 separate regressions. Moreover,  
544 concern about utility-level spillovers motivated creation of comparison samples from within  
545 Zarqa (and subject to utility water supply re-optimization), and from neighboring areas in East  
546 Amman (that would not benefit from such spillovers).

547 The farm comparison is somewhat different, in that separate regressions are estimated to  
548 compare outcomes in each of five survey areas – selected on the basis of their varying levels of  
549 baseline and Compact-induced exposure to treated wastewater – to those in all other zones.  
550 These five areas include four different zones in the Jordan Valley (JV) as well as an additional  
551 zone located along the Zarqa River (Figure 3). In this regression, we are especially interested in  
552 changes in the mid-north of the JV (labelled zone JV2), where treated effluent as well as surface  
553 waters blended with treated effluent was introduced to irrigators for the first time during this  
554 period [*Morgan et al.*, 2021]. Impacts measured in this regression are not strictly limited to the

555 impact of the JC, but rather represent the collective effects of several simultaneous changes in  
556 irrigation water sourcing by farmers.<sup>5</sup>

557 The outcomes  $Y_{ijt}$  that we consider are the main factors that were expected – based on the full  
558 program theory of change – to evolve in the short to medium term, due to the JC investments.  
559 Among households, we focus on measures of the reliability of water supply, billed water  
560 consumption, perceptions of the quality of service (water pressure and safety), service  
561 disruptions and sewer backups, expenses for non-network water purchases, sewer connections,  
562 and pit-emptying costs. For firms, we consider changes in network water consumption and  
563 connections, sewerage, water expenses, and net profitability. Among farmers, we analyze water  
564 sourcing, perceptions of water quality, decisions about cropping across seasons, revenues, and  
565 profits.

566 We also note that the estimated parameter  $\kappa$  in equation 1 represents an intention-to-treat (ITT)  
567 estimate; that is, it measures impacts on households and farmers whether they choose to comply  
568 with the intervention or not (Galasso et al. 2004). This is also the most relevant policy parameter,  
569 because it accounts for the behavioral responses of all populations exposed to the JC, whether or  
570 not they choose to take up specific infrastructures or improvements. For example, some  
571 households may not connect to new sewer pipes, some may not consume additional network  
572 water, and some farmers may reject blended wastewater in favor of groundwater or other  
573 alternatives.

574 A key threat to clean identification of impacts using any DiD approach is that time-varying  
575 unobserved differences across groups may be responsible for differential changes that are  
576 observed, rather than the JC investments *per se*. In the household survey, this threat is mitigated  
577 by matching households located in areas with improvements to households outside of those  
578 areas, which reduced *ex ante* differences within the sample, and by the verification of parallel  
579 pre-intervention trends in network water consumption across treated and comparison zones (see  
580 Appendix 1 for details). Due to data limitations, we are unable to establish parallel trends among  
581 farms in the JV, however, and emphasize that the natural experiment we exploit – driven by

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<sup>5</sup> These changes included other complementary infrastructure works, including most notably the establishment of a new connector between the King Talal Dam that stores blended Zarqa River flow and treated wastewater, and the mid-north JV2 zone. They also include general increases throughout the region, in the availability of treated wastewater.

582 water system managers' replacement of freshwater supply to farmers with treated effluent – had  
583 already been underway for some time prior to the JC, per the GoJ policy encouraging that water  
584 substitution. In zone 2, however, the natural experiment we exploit does pertain to the new  
585 introduction of treated wastewater in irrigation, since this zone had not been connected to treated  
586 effluent prior to the JC. This is important for interpreting the trends we observe in the farm  
587 sample.

588 The other substantial element in the analysis of JC impacts is a water balance modeling exercise,  
589 parameterized based on comparative tracking of utility performance indicators over time, and  
590 from qualitative interviews with key stakeholders, as mentioned in Section 4.3 above. In the first  
591 of these, we modify an existing Jordan-wide water allocation model built using the Water  
592 Evaluation and Planning (WEAP) System software used by planners in the Ministry of Water  
593 and Irrigation (MWI) of the GoJ. Specifically, we create “with JC” and “without JC” scenarios  
594 that differ only in the following two parameters: a) the physical loss in water supply to  
595 beneficiaries in Zarqa Governorate; and b) the sewerage rate. We then specify these parameters  
596 based on the best empirical estimates of how these factors changed after JC implementation. This  
597 water balance analysis allows us to track systems-level changes in water supply to various users  
598 – irrigators in the Jordan Valley, urban consumers in Zarqa, and urban consumers elsewhere in  
599 Jordan, owing to the primary substitution mechanism, that serve to further contextualize and  
600 validate our IE estimates and the program theory of change. More details on the structure and  
601 assumptions of the water balance analysis can be found in Jeuland et al. [2021a].

602

## 603 **5. Evaluation implementation and results**

604 The multi-pronged evaluation strategy described in Section 4 provides evidence of several, but  
605 not all, of the changes expected from the Compact, and thereby provides only partial support for  
606 the originally hypothesized theory of change. Below, we summarize in succession the main  
607 results among beneficiaries in Zarqa (households and small enterprises), farmers in the Jordan  
608 Valley, and finally discuss the relative utility performance measures. Additional details can be  
609 found in Jeuland et al. [2020b].

610 *Impacts on households and small businesses in Zarqa*

611 In areas exposed to the Water Network Project, there is strong evidence of improvements in  
612 households' reporting of water pressure, increased network water use, and reduced complaints  
613 about water shortage, as well as somewhat weaker evidence of increases in the hours of water  
614 received from the piped network (Table 3, columns labelled "WNP" and "Both"). The  
615 specifications presented in Table 3 control for time and household fixed effects.<sup>6</sup> These positive  
616 changes notwithstanding, a key category of anticipated impacts that would increase net  
617 household income – reduced spending on expensive alternatives to utility water and on water  
618 overall – does not appear to have materialized.<sup>7</sup> There are several potential explanations for the  
619 lack of cost savings among households. First, the evaluation may have been underpowered to  
620 detect such impacts, since vendor purchases varied widely in the sample (notably, most of the  
621 relevant coefficient estimates, ranging from 5 to about 10 JD/month are negative but imprecisely  
622 measured, i.e., they are not statistically significant at conventionally reported levels). Perhaps  
623 more importantly, though, perceptions of the quality of networked water did not change  
624 following the investment. This, coupled with data showing that customers have generally  
625 negative perceptions of the quality of networked water relative to vendor water, seems to have  
626 manifested in a continuing preference for purchasing treated water containers (generally in 20L  
627 jugs) among many households, at least for drinking purposes.<sup>8</sup> Finally, the lack of clear  
628 substitution away from treated vendor water could have been due to an insufficient increase in  
629 the reliability of water supply to warrant shifting away from non-network alternatives; after all,  
630 supply remained intermittent. This highlights the importance of continuous water supply.  
631 Consistent with the apparent lack of substantial household substitution, then, there was no  
632 evidence of households saving time on water collection or procurement, following the  
633 intervention (results not shown).

634 Considering next the areas exposed to the Wastewater Network Project, there is strong evidence  
635 of an increase in households' likelihood of being connected to the piped sewer network and  
636 lower use of stand-alone cesspits, but only weak evidence of cost savings on emptying septic

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<sup>6</sup> Alternative specifications that additionally control for two key time-varying factors – the number of refugees arriving in the sample area (a water demand shock) and a household-specific wealth index – yield very similar results. These alternative results are available upon request from the authors.

<sup>7</sup> This is also confirmed by the lack of evidence in the vendor survey for a decline in these activities over the period of the JC.

<sup>8</sup> Households reported that their stored drinking water – often purchased from shops – was safer on average than utility water. This perception is at odds with our own test results for *e. coli* and *total coliform*, which showed that network water from household taps contained *lower* contamination than stored drinking water, on average.

637 tanks, and no real evidence of reduced sewer backups, which are nonetheless infrequent (Table  
638 3, columns labelled “WWNP” and “Both”). The cost savings outcome may take longer to  
639 manifest since these tanks take time to fill up, or else the evaluation may have been  
640 underpowered to detect it (we note that coefficients are negative in magnitude but imprecisely  
641 estimated). Finally, and in line with findings that Zarqa residents exposed to the improvement  
642 did not reduce water expenditures or shift away from expensive vendor sources, and that cost  
643 savings on septic tank emptying were unclear, there is no consistent evidence that the JC led to  
644 economic improvements among households treated by the WNP, WWNP, or the combination of  
645 these interventions (in terms of income, expenditure, assets). If anything, net income might have  
646 decreased slightly, perhaps due to the one-time cost of connecting to the sewer network.

647 Finally, there is some support for the idea that the investment generated positive spillovers  
648 within Zarqa, compared to neighboring areas in Amman supplied by a different water utility.  
649 This is apparent in the relatively larger impacts detected in the comparison with controls in  
650 Amman, for several key water supply variables: reported water pressure, hours of supply on days  
651 with water, and stronger reductions in water shortages. Impact estimates for the water supply  
652 improvements based on the Zarqa comparisons, therefore, appear biased downward due to  
653 positive spillovers. We would not expect similar spillovers through the wastewater  
654 improvements on most measured outcomes, except perhaps for sewer backups, but no impact on  
655 that outcome was detected. Consistent with this expectation, there do not appear to be  
656 consistently larger impacts relative to control households in Amman, for use of stand-alone  
657 wastewater systems, connection to the sewer, or savings on emptying septic tanks.

658 In contrast to the generally positive impacts observed on households due to the investments,  
659 enterprises did not appear to experience a similar increase in water consumption or service  
660 reliability. This is likely because rates of connection to the piped network remained low among  
661 this population (Table 4), and promoting connections among firms – which tend to be much  
662 higher in cost than household connections – was not an explicit objective of the Compact. In any  
663 case, it does not appear to be due to lack of statistical power, since changes for nearly all  
664 outcomes are small or even negative. Thus, enterprises did not appear to benefit the way  
665 households did from this investment.

666

667 *Impacts on farmers in the Jordan Valley*

668 Among farmers in Compact-affected areas, we observe increased supply and use of blended  
669 wastewater for irrigation, relative to areas that were outside the Compact-affected areas (Table  
670 5). The areas most impacted by such increased flows are the JV2 and JV3 locations, followed by  
671 those in JV1 (which were newly receiving treated effluents from sources near Irbid) and the  
672 highland farms located along the Zarqa River. This greater water availability in turn mostly led  
673 to increased land area being irrigated, but was considered to reduce water quality in the areas  
674 receiving treated wastewater for the first time – JV1 and JV2. Perhaps owing to these water  
675 quality impacts (particularly the increased salinity of treated wastewater), the relative increase in  
676 water availability did not clearly translate into changes in the overall value of farm output or  
677 profits. The one exception to this pattern was in the highlands, where impacts were large and  
678 positive, and also seen for assets. The outcomes for highland famers suggests that irrigators  
679 located upstream of the Jordan Valley may have captured many of the expected benefits of the  
680 enhanced irrigation water availability, which had not been anticipated by Compact planners.

681 Meanwhile, further downstream in affected areas of the Jordan Valley, relative vegetable  
682 production and farm input costs increased as tree output decreased, suggesting a substitution  
683 away from saline- (or wastewater-) intolerant tree crops and into less sensitive horticulture, as  
684 shown in other work [*Morgan et al.*, 2021]. Finally, it may be reasonable to infer, based on the  
685 water balance analysis [*Jeuland et al.*, 2021a] and the fact that land values in these areas  
686 remained stable, that the investment increased the value of farm output relative to a no-Compact  
687 counterfactual with increasing water scarcity, in which agricultural activity would not have been  
688 similarly sustained.

689

690 *Evolution in the behavior and performance of the utility*

691 Comparing performance measures for the Zarqa water utility against other corporatized urban  
692 utilities in Jordan (namely in Amman and Aqaba), the JC appears to have improved measures of  
693 utility functionality. This elevated performance is reflected in evidence of sharply lower  
694 incidence of pipe failures (after an initial uptick in leaks during the infrastructure transition),  
695 declining administrative losses as measured by billing and collection efficiency, higher utility

696 revenue collection, and reduced overall non-revenue water (NRW) measured in volume per  
697 subscriber per day (Table 6). As noted previously, the WNP component of the project primarily  
698 aimed to reduce physical water losses, but may also have reduced administrative losses through  
699 meter replacement. Meter testing on a small sample of meters in survey areas conducted prior to  
700 replacements indicated 25% under-measurement of consumption on average (results available  
701 upon request from the authors), and showed increased billed consumption following replacement  
702 in both rehabilitated and non-rehabilitated network areas (likely due to more accurate  
703 measurement of consumption). It is well known that the accuracy of mechanical meters  
704 deteriorates in intermittent systems, as a function of age, water pressure fluctuations, and the  
705 presence of air in the network [*Walter et al.*, 2017].

706 That said, the evaluation cannot differentiate the effect of the JC investments from that of the  
707 contemporaneous utility corporatization reform, which likely also affected indicators of  
708 functionality and efficiency. Notably, however, NRW declines resulting from the intervention  
709 did underperform relative to JC targets and expectations, especially when measured in  
710 percentage terms. This may have been driven by several factors including incomplete isolation of  
711 rehabilitated network areas, increases in water supply among portions of the network outside the  
712 intervention area (or spillovers from greater water allocation by managers, to better meet demand  
713 in unimproved water scarce areas), or illicit water use. Key informant interviews with utility  
714 personnel helped to support these explanations; several senior utility operators noted that water  
715 “rotations” – periods of service to different neighborhoods within Zarqa – were adjusted  
716 following the Compact to allow the benefits of the improvements to be shared more equally  
717 across households in improved and unimproved areas. Another area of potential spillovers was in  
718 maintenance; because servicing needs declined in WNP areas, effort could be reallocated to non-  
719 Compact areas [*Jeuland et al.*, 2020b].

720 Finally, rising operating costs meant that short-term improvements in the Operating Cost  
721 Recovery Ratio between 2014 and 2016 were not sustained. Higher operating costs were largely  
722 driven by increased per-unit volume energy costs and the cost of additional wastewater  
723 management, as well as increase of imported water pumped from the far-away Disi aquifer.

724

## 725 **6. Discussion and conclusions**

726 This paper provides a contemporary discussion of the problem of impact evaluations of large  
727 water infrastructure projects. The contemporary perspective is important in light of concerns  
728 over the validity and plausibility of previous assessments [*Angrist and Pischke*, 2010], the  
729 particular challenges of these long-lived projects, the complex causal chains that determine their  
730 impacts [*Polasky et al.*, 2019; *Tallis et al.*, 2019], and the relatively limited evidence on which to  
731 base future similar investment decisions. To that end, we described the core challenge facing  
732 researchers aiming to establish the causal nature of observed changes in outcomes, and then  
733 proposed a set of three key criteria – validity, relevance and practicality – that researchers  
734 working in this domain should consider when designing and implementing evaluation research.  
735 As there are tradeoffs across these three criteria, scholars must thoughtfully weigh the particular  
736 pros and cons, and opportunities and vulnerabilities of different methods, to advance knowledge  
737 and achieve greater policy impact.

738 We demonstrated the approach with discussion of a multi-faceted evaluation implemented to  
739 assess the outcomes attributable to the Jordan Compact, a highly integrated project that  
740 combined water and wastewater investments, with the goal of reducing water scarcity and  
741 enhancing economic opportunity in Zarqa, Jordan. Overall, the evaluation found a positive net  
742 economic impact, though impacts in some domains fell short of expectations [*Jeuland et al.*,  
743 2020b]. On the positive side, there was clear evidence of improved service and reduced water  
744 loss where the water network was rehabilitated. Investments in sewerage meanwhile increased  
745 the number of household connections and reduced the incidence of sewer backups. Finally, the  
746 water savings and increased wastewater capture in Zarqa led to increased supply of treated  
747 wastewater to irrigators along the Zarqa River and in the Jordan Valley, which is supporting a  
748 profit-neutral shift away from cultivation of water-intensive and salinity-sensitive citrus trees,  
749 and toward vegetable field crops. This shifting supply, in turn, is freeing up freshwater  
750 previously allocated to irrigation for higher-value uses in urban areas.

751 On the negative side, however, there was no evidence that the JC improvements reduced  
752 expenses for high cost non-network water, which was a key expected channel of benefits to  
753 households. Such behavioral changes may take longer to manifest, but survey evidence suggests  
754 that households maintained their skepticism about the safety of piped network water (*Orgill-  
755 Meyer et al.* 2018). Moreover, small and medium enterprises did not benefit from the

756 investments, likely owing to their continued low rate of connection to the piped water and sewer  
757 system. Finally, improvements in water supply reliability to rehabilitated areas were somewhat  
758 diminished by utility adjustments that increased supply to areas outside the intervention area.  
759 Ultimately, the mixed realized benefits of the investments in Zarqa may therefore perpetuate a  
760 low-equilibrium trap that plagues water utilities in developing countries and is difficult to resolve  
761 sustainably [Jeuland, 2022]. Specifically, because consumers do not trust water utilities, they  
762 often resist paying tariffs that allow full cost-recovery, or engage in theft from the water  
763 network. Such behaviors compromise the utility's ability to invest in long-term maintenance and  
764 reliability.

765 Overall, the evaluation contributes to a relatively thin literature on the economic benefits of  
766 investments in urban water and sewer systems, which represent one of the most important quasi-  
767 public goods provided by governments in low- and middle-income countries. Examined through  
768 the lens of the economic analysis justifying the JC, the success of the intervention was largely  
769 contingent on the effective substitution, for irrigation uses, of recycled wastewater for freshwater  
770 supplies. The conditions leading to successful substitution of the type observed in this case have  
771 rarely been documented [Jeuland, 2012], and this represents an important new contribution of  
772 this evaluation to the literature. Indeed, most studies in the Middle East and globally have rather  
773 highlighted the relatively limited success of attempts to increase wastewater reuse owing to  
774 salinity or other water quality concerns [Carr *et al.*, 2011; Jeuland, 2015].

775 Perhaps more significantly, though, the JC example serves as a useful demonstration of a number  
776 of evaluation issues and challenges discussed in this paper, that warrant additional research and  
777 policy engagement. First, researchers seeking to carry out policy-relevant evaluations of large  
778 infrastructure investments must work harder to engage with project planners to understand these  
779 interventions' complete theories of change and to track the most important set of anticipated  
780 impacts. The eventual evaluation design in this case, with data analysis focused on a range of  
781 stakeholder groups and both pecuniary and non-pecuniary quality of life outcomes, could then be  
782 crafted to allow a critical appraisal of planners' most critical assumptions, rather than focusing  
783 on a narrow set of questions (e.g., whether the investment reduced diarrheal disease incidence).  
784 Second, and relatedly, it allowed for nuanced understanding of the distributional effects of the  
785 investment. Specific and documented failures, for example related to water consumers lack of

786 confidence in utility water in this case, can then inform development of future remedies to  
787 address them. Thus, Zarqa policy-makers might consider investing to convince users of the  
788 safety of network water, given our results showing that this water may be safer than more  
789 expensive water purchased from vendors [*Orgill-Meyer et al.*, 2018].

790 Third, large infrastructure projects nearly always have spillovers and overlapping or systems-  
791 level impacts, which good evaluations must try to anticipate. The JC project is a particularly  
792 salient example of this, with its highly integrated design. Owing to a fairly sophisticated  
793 understanding of the project theory of change, the evaluation therefore worked to combine  
794 several complementary data collection and quasi-experimental analytical techniques to provide a  
795 comprehensive view of the investment impacts. Importantly, this also motivated construction of  
796 two alternative comparison samples (one from unimproved areas in Zarqa, which were highly  
797 subject to infrastructure spillovers but highly comparable to treated areas, and one from  
798 neighboring areas in Amman Governorate, which were not subject to spillovers but also less  
799 comparable). This approach allowed for learning about both spillovers and impacts, which is  
800 highly valuable for policy-making.

801 Fourth, there are often important tradeoffs between theoretical internal validity (with RCTs  
802 serving as a gold standard) and risks of contamination of the evaluation, which occurs when  
803 areas identified *ex ante* for treatment are not improved, or when evaluation comparison areas end  
804 up receiving improvements. With the JC, such threats were borne out, as one planned  
805 neighborhood ended up not being rehabilitated due to local politics, and Compact implementers  
806 seeking to exhaust the implementation budget also worked to extend the improvements beyond  
807 the originally planned areas. Throughout the process, implementer-evaluator coordination helped  
808 to minimize the risks of contamination, and clarify their severity, such that the design's integrity  
809 was ultimately maintained. Upon reflection, much of the success for this was due to the  
810 communication and trust between the two parties, and stemmed from implementers' appreciation  
811 for the evaluation's efforts to create a cost-effective and pragmatic evaluation that was respectful  
812 of the original project design and objectives.

813 Finally, a critical limitation of this specific evaluation that is highly relevant in the context of  
814 infrastructure projects is its relatively short, four-year time frame (from baseline to endline).  
815 Indeed, a focus on short-run and medium-term impacts is a challenge to IE in infrastructure

816 domains that may take many years to realize benefits. Even as there is also substantial  
817 uncertainty about the long-term performance of such investments and the continued evolution of  
818 beneficiary behavior, however, which should motivate longer-term work, the validity of the  
819 treatment-control comparisons become increasingly tenuous as time goes on. Future research on  
820 the JC's effects should attempt to verify the persistence of the short-term changes measured here,  
821 and the project's distributional consequences. For example, farmers in the Northern Jordan  
822 Valley will likely continue to adjust to the shift in water supply over time, and households and  
823 businesses may gain confidence in the quality of network water or persist in purchasing more  
824 expensive alternatives. Data from Miyahuna-Zarqa could help to obtain a more complete picture  
825 of the effects on the utility, and whether short-term changes in NRW continue or are reversed.

826

827

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## Tables and figures

**Table 1.** Summary of evaluation options, with focus on main internal validity threats, relevance, and practical considerations that are of particular importance for network water supply and sanitation

Method	Description and comments	Threats to validity of causal inference	Relevance of evaluation evidence	Practical / logistical considerations
<u>Experimental</u>				
Randomized Controlled Trial (RCT) [Duflo et al., 2007]	RCTs are generally not feasible for network water infrastructure, as such interventions are clustered, directional, and designed to serve population at scale or to address known (selected) system deficiencies. Some complementary interventions (information campaigns) can be evaluated using this approach. Smaller-scale rural infrastructure (e.g., condominal sewerage, village-scale piped water) can be evaluated with cluster RCTs, or step-wedge RCTs.	<ul style="list-style-type: none"> <li>• <i>Confounding</i> due to unbalanced randomization</li> <li>• <i>Spillovers</i> (violation of the stable unit treatment value assumption, or SUTVA), whereby some units benefit as a result of other units' uptake.</li> <li>• Vulnerable to <i>selective attrition</i></li> </ul>	<ul style="list-style-type: none"> <li>• Typically <i>artefactual</i>, w/ limited evaluation questions</li> <li>• Treatment effect can be <i>representative</i></li> <li>• “Gold standard” for causal researchers</li> <li>• Results are <i>not conditioned</i> by assumptions</li> <li>• <i>Statistical power</i> is a design feature, but usually sufficient for a few pre-identified outcomes</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: High, especially when powered for multiple outcomes or interventions</li> <li>• <i>Contamination risk</i>: Moderate, as pressure to help “untreated” units increases over time</li> <li>• <i>Coordination</i>: Mainly pertains to maintaining integrity of randomization</li> <li>• <i>Interpretation</i>: Intuitive and highly transparent</li> <li>• <i>Pre-intervention data needs</i>: Low to none</li> <li>• <i>Flexibility to adapt</i>: Very low</li> </ul>
Experimental encouragement design [Katz et al., 2001]	Subsidies or other assistance to customers can generate exogenous variation in the take-up of infrastructure connections, for use as an instrumental variable for isolating impacts. The resulting local average treatment effect is specific to those who respond to the encouragement [Heckman et al., 2006].	<ul style="list-style-type: none"> <li>• Same as above</li> </ul>	<ul style="list-style-type: none"> <li>• Same as above, except that the treatment effect only applies to the population that responds to the encouragement</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low to moderate, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Low</li> <li>• <i>Coordination</i>: Moderate; mainly in combining with other methods (DiD) to strengthen validity</li> <li>• <i>Interpretation</i>: Intuitive but not always transparent</li> <li>• <i>Pre-intervention data needs</i>: Low to none</li> <li>• <i>Flexibility to adapt</i>: Impossible</li> </ul>
<u>Quasi-experimental</u>				
Natural experiment [J Angrist et al., 2002]	Some infrastructure placements are determined by geographic or other factors that are “as good as random” in determining exposure to improvements, such that they provide researchers with “natural experiments” [Cerdá et al., 2012], that give rise to comparable treatment and control groups. Another version of this is an interrupted time series analysis where a time-dependent event (e.g., rehab of one part of a water network) gives rise to a sharp change that only affects some households or others.	<ul style="list-style-type: none"> <li>• <i>Confounding</i> by geographic / other factors determining exposure may also confound outcomes</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA) outside of treatment area</li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect is <i>representative</i> but contingent on natural experiment conditions</li> <li>• Generally accepted by researchers</li> <li>• Results are <i>not conditioned</i> by assumptions</li> <li>• <i>Statistical power</i>: Difficult to</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low to moderate, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Low</li> <li>• <i>Coordination</i>: Moderate; mainly in combining with other methods (DiD) to strengthen validity</li> <li>• <i>Interpretation</i>: Intuitive but not always transparent</li> <li>• <i>Pre-intervention data needs</i>: Low to none</li> <li>• <i>Flexibility to adapt</i>: Impossible</li> </ul>

			anticipate ex ante	<ul style="list-style-type: none"> <li>• <i>Other</i>: Natural experiment can be hard to anticipate</li> </ul>
Difference-in-differences (DiD) [Card and Krueger, 2000]	In this approach, impacts are estimated by subtracting out the trend in an unexposed sample, which represents the counterfactual, from that in an exposed sample. Such samples are created using variation in spatial targeting or other eligibility criteria, which are common for network water infrastructure extension or rehabilitation. The validity of the comparison relies on pre-treatment trends being similar in the groups, and can be enhanced using matching or econometric models that control for differences in baseline covariates.	<ul style="list-style-type: none"> <li>• <i>Confounding</i> by time-varying unobservables</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA)</li> <li>• Vulnerable to <i>selective attrition</i></li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect is usually <i>representative</i> (unless combined w/other methods)</li> <li>• Generally accepted by researchers, subject to showing parallel trends</li> <li>• Results are <i>not conditioned</i> by assumptions</li> <li>• <i>Statistical power</i> is a design feature</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Moderate to high, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Moderate to high</li> <li>• <i>Coordination</i>: Moderate; mainly in combining with other methods (matching) to strengthen validity</li> <li>• <i>Interpretation</i>: Intuitive and transparent</li> <li>• <i>Pre-intervention data needs</i>: Moderate to high (parallel trends)</li> <li>• <i>Flexibility to adapt</i>: Moderate</li> </ul>
Matching or synthetic control [Abadie and Gardeazabal, 2003; Rosenbaum and Rubin, 1985]	These methods are best when combined with DiD analysis, but can be used to improve comparability when targeting is correlated with baseline characteristics. Various matching approaches enhance comparability by sampling untreated observations that can approximate the treatment counterfactual. For example, propensity score matching (PSM) finds treated and untreated observations that have a similar probability of being treated, from a regression of participation on observables. Synthetic control uses a time series of pre-intervention observations to “train” an algorithm that identifies weights for a pool of observations with similar counterfactual trends as one or more treated units.	<ul style="list-style-type: none"> <li>• <i>Confounding</i> by unobservables (Conditional Independence Assumption), worse when match quality is low</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA)</li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect only applies to units with suitable comparisons (common support region)</li> <li>• Researchers are often skeptical that the CIA has been met</li> <li>• Results are <i>conditioned</i> by assumptions of the matching algorithm</li> <li>• <i>Statistical power</i> is a design feature</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Moderate to high, depending on data collection needs</li> <li>• <i>Contamination risk</i>: High</li> <li>• <i>Coordination</i>: Moderate; mainly in combining with other methods (DiD) to strengthen validity</li> <li>• <i>Interpretation</i>: Intuitive, but matching may lack transparency</li> <li>• <i>Pre-intervention data needs</i>: Moderate (matching)</li> <li>• <i>Flexibility to adapt</i>: Moderate</li> </ul>
Instrumental variables (IV) [J D Angrist and Krueger, 2001]	An instrumental variable is a factor that predicts exposure to or participation in an intervention, but that does not affect outcomes directly through channels other than that effect on participation. This creates exogenous variation in the intervention that can be leveraged to determine its impacts. The impact measure is a local average treatment effect that measures the effect of the intervention on those (“compliers”) whose participation is affected by the instrument. Program placement rules or constraints may give rise to valid instruments.	<ul style="list-style-type: none"> <li>• <i>Confounding</i>: For many interventions and outcomes, there are few plausibly “exogenous” assignments of this type, at least in a statistical sense</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA)</li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect (LATE) is not representative, and not always for the most relevant population</li> <li>• Researchers are often skeptical about exclusion restriction</li> <li>• Results are <i>conditioned</i> by exogeneity assumptions</li> <li>• <i>Statistical power</i> is often</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low to moderate, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Not applicable</li> <li>• <i>Coordination</i>: Low</li> <li>• <i>Interpretation</i>: Unintuitive, lacks transparency</li> <li>• <i>Pre-intervention data needs</i>: Low</li> <li>• <i>Flexibility to adapt</i>: High</li> <li>• <i>Other</i>: Suitable IV may not exist</li> </ul>

			reduced by 2-stage estimation	
Regression discontinuity (RD) [Imbens and Lemieux, 2008; Thistlethwaite and Campbell, 1960]	RD exploits discontinuities in eligibility for an intervention with respect to an assignment variable. For example, population thresholds, or a poverty line threshold for subsidy eligibility.	<ul style="list-style-type: none"> <li>• <i>Confounding</i>: Eligibility rule violations or manipulation, or “fuzzy” discontinuities that are difficult to characterize well</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA)</li> <li>• Vulnerable to <i>selective attrition</i></li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect is limited to units very near the discontinuity</li> <li>• Generally accepted by researchers</li> <li>• Results are <i>conditioned</i> on proximity to eligibility cutoff</li> <li>• Statistical power may be limited</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low to moderate, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Moderate, depending on rigor with which eligibility is assessed</li> <li>• <i>Coordination</i>: Low</li> <li>• <i>Interpretation</i>: Intuitive, but transparency may be lacking due to definition of the RD bandwidth</li> <li>• <i>Pre-intervention data needs</i>: Low</li> <li>• <i>Flexibility to adapt</i>: Low</li> </ul>
<u>Other</u>				
<i>Ex post</i> regression	Statistical comparison of treated and untreated units, with statistical control for observed differences between the groups. Also commonly called “observational” comparisons.	<ul style="list-style-type: none"> <li>• <i>Selection</i>: Units that participate are systematically different than those that do not</li> <li>• <i>Confounding</i> by unobservables</li> <li>• <i>Spillovers</i> (i.e., violation of SUTVA)</li> </ul>	<ul style="list-style-type: none"> <li>• Evidence arises directly from the <i>real world</i></li> <li>• Treatment effect is usually representative</li> <li>• Causal researchers are typically highly skeptical of results</li> <li>• Results are <i>conditioned</i> on controls</li> <li>• <i>Statistical power</i>: Difficult to anticipate ex ante</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low to moderate, depending on data collection needs</li> <li>• <i>Contamination risk</i>: Not applicable</li> <li>• <i>Coordination</i>: Low</li> <li>• <i>Interpretation</i>: Intuitive, but transparency may be lacking (contingent on choice of controls)</li> <li>• <i>Pre-intervention data needs</i>: None</li> <li>• <i>Flexibility to adapt</i>: High</li> </ul>
Counterfactual modeling [Balke and Pearl, 2013]	Complex water resources systems evolve stochastically according to both human and environmental influences. This approach leverages systems understanding from socio-hydrological or hydro-economic models to conduct “with” and “without” simulations of interventions, for construction of model-based comparisons [Srinivasan, 2015].	<ul style="list-style-type: none"> <li>• <i>Confounding</i> by behavioral or other system-level factors not accounted for</li> </ul>	<ul style="list-style-type: none"> <li>• Evidence is <i>artefactual</i>; model may diverge from real world observations</li> <li>• Treatment effect is usually representative, but may not align with policy-maker priorities and needs</li> <li>• Not widely used by causal social science researchers, who are wary of over-calibration</li> <li>• Results are <i>conditioned</i> on model assumptions</li> <li>• <i>Statistical power</i>: Not applicable</li> </ul>	<ul style="list-style-type: none"> <li>• <i>Cost</i>: Low</li> <li>• <i>Contamination risk</i>: Not applicable</li> <li>• <i>Coordination</i>: Low</li> <li>• <i>Interpretation</i>: Not intuitive and not always transparent (requires interdisciplinary expertise)</li> <li>• <i>Pre-intervention data needs</i>: Moderate to high, depending on calibration needs</li> <li>• <i>Flexibility to adapt</i>: High</li> <li>• <i>Other</i>: Required model effort is substantial</li> </ul>

**Table 2.** Summary of study populations and data collection methods deployed

Survey element	Survey type	Sampling frame	Sample selection	Stratification / comparison group	Representation	Sample size
Household	4-wave Panel	Zarqa and Amman (from Jordan Dept. of Statistics (DoS))	<ul style="list-style-type: none"> <li>Survey geocodes selected based on <i>ex ante</i> matching of treated and control zones, using Census data</li> <li>Random sampling within sample geocodes</li> <li>Replacements selected from sample geocodes</li> </ul>	WNP only – WNP only control WWNP only – WWNP only control WNP+WWNP – WNP+WWNP control  Distinct control groups from: <ul style="list-style-type: none"> <li>Zarqa</li> <li>Amman</li> </ul>	Representative of sample geocodes at baseline, based on comparisons to Census and other sources	1. 3359 2. 3416 3. 3596 4. 3662
Enterprise	2-wave Panel	Zarqa and Amman (from DoS + household referrals)	<ul style="list-style-type: none"> <li>Same geocodes as household sample</li> <li>Random selection within sample geocodes</li> <li>Referrals for informal enterprises</li> <li>Replacements selected from closely neighboring enterprises</li> </ul>	Same as household (though analysis uses all controls for each group to maximize statistical power)	Representative of sample geocodes at baseline Informal enterprises likely under-represented (due to low referral rates)	1. 345 2. 418
Farm	3-wave Panel	Jordan Valley and highlands (from DoS)	<ul style="list-style-type: none"> <li>Survey zones selected based on expected differences in exposure to treated wastewater</li> <li>Random selection in sample zones</li> <li>Replacements selected within zones</li> </ul>	Five locations: <ul style="list-style-type: none"> <li>Highlands u/s KTD (↑ river flow)</li> <li>JV1 North (↑ Non-Compact WW)</li> <li>JV2 Mid-North (↑ Compact WW)</li> <li>JV3 North-Central (↑ Compact WW)</li> <li>JV4 South-Central (little change in WW)</li> </ul>	Representative of sample zones	1. 551 2. 539 3. 539
Refugee	Single cross-section	UNHCR registration list for Zarqa and Amman	<ul style="list-style-type: none"> <li>Priority survey geocodes selected according to household sample, with augmenting based on treatment status outside hh geocodes</li> <li>Random sampling by treatment status</li> <li>Referrals for unregistered refugees</li> </ul>	Treatment status: <ul style="list-style-type: none"> <li>WNP only</li> <li>WWNP only</li> <li>Both WNP and WWNP</li> <li>Controls in Zarqa</li> <li>Control Amman</li> </ul>	Representative of registered population in sample areas Unregistered population likely under-represented (due to low referral rates)	1617
Water vendor	Single cross-section	Shops: Ministry of Health list + canvassing Tankers: Canvassing	<ul style="list-style-type: none"> <li>Full sampling from canvassed locations</li> </ul>	None	Representative of water vendors in Zarqa and East Amman in 2018	320
Meter testing	Repeat cross-section	Meter listing in selected zones	<ul style="list-style-type: none"> <li>Zones selected for variation in JC status, elevation, pressure, throughput (for survey 1), and JC status and meter replacement (for survey 2)</li> <li>Random sample of meters within selected zones</li> </ul>	Compact and non-Compact zones	Not representative	1. 37 2. 223
Water loss testing	Single cross-section	Canvassing of land plots in selected areas	<ul style="list-style-type: none"> <li>“Well isolated” zones selected (as suggested by utility)</li> <li>Comparison of meter registered data to bulk meter inflow</li> <li>Random sub-sample of meters evaluated to adjust for meter error</li> </ul>	Meter error testing sub-sample stratified by meter replacement status	Not representative; only relevant to “well isolated” zones	1797
Key informant interviews	Single cross-section	Listing of key JC stakeholders	<ul style="list-style-type: none"> <li>Contact to all listed stakeholders</li> <li>Replacements included as suggested by stakeholders</li> </ul>	None	Representative of institutions, but likely not all perspectives	22

**Table 3.** Summary of main impacts on household behaviors and outcomes

Outcome	DiD impact of intervention – relative to non-intervention areas in Zarqa, by subsample			DiD impact of intervention – relative to non-intervention areas in Amman, by subsample		
	(1) WNP	(2) WWNP	(3) Both	(4) WNP	(5) WWNP	(6) Both
<u>Water supply</u>						
Reported water pressure rating <sup>1</sup>	<b>-0.38***</b> (0.15)	n.a.	<b>-0.49***</b> (0.14)	<b>-0.63***</b> (0.12)	n.a.	<b>-0.84***</b> (0.14)
Reported perception of network water quality	<b>+0.63**</b> (0.30)	n.a.	-0.29 (0.34)	+0.32 (0.27)	n.a.	-0.58 (0.36)
Assessed water quality (E. coli count) <sup>2</sup>	-0.049 (0.053)	n.a.	0 (0)	n.a.	n.a.	n.a.
Hours piped water, for days w/water	+0.86 (0.61)	n.a.	+0.37 (0.64)	<b>+1.15**</b> (0.46)	n.a.	<b>+1.83***</b> (0.47)
Reported water shortage, past month	<b>-0.10*</b> (0.05)	n.a.	-0.05 (0.06)	<b>-0.12***</b> (0.04)	n.a.	<b>-0.098*</b> (0.05)
Network water use – Utility sample <sup>3</sup>	<b>+2.9***</b> (0.66)	n.a.	<b>+2.9*</b> (1.5)	+0.52 (0.59)	n.a.	<b>+1.9*</b> (1.1)
Network water use – Survey sample		n.a.	+6.1 (3.9)	+5.1 (3.5)	n.a.	+3.7 (4.3)
Expenditure on water from vendors (JD/month)	-5.1 (5.0)	n.a.	-7.1 (5.5)	-6.8 (4.4)	n.a.	<b>-9.7*</b> (5.8)
Expenditure on water, all sources (JD/month)	-3.6 (5.8)	n.a.	-6.2 (5.7)	-4.0 (5.1)	n.a.	-10.6 (6.3)
<u>Wastewater management</u>						
Use of stand-alone cesspits	n.a.	<b>-0.13***</b> (0.04)	<b>-0.07*</b> (0.04)	n.a.	<b>-0.14***</b> (0.05)	<b>-0.11*</b> (0.04)
Sewer connection	n.a.	<b>+0.14***</b> (0.05)	<b>+0.17***</b> (0.05)	n.a.	<b>+0.09*</b> (0.05)	<b>+0.12**</b> (0.05)
Expense for septic tank evacuation	n.a.	-1.0 (4.7)	-15.1 (11.3)	n.a.	-7.7 (5.6)	-18.2 (11.2)
Sewer backup prevalence	n.a.	-0.02 (0.01)	-0.004 (0.01)	n.a.	<b>-0.02**</b> (0.01)	-0.002 (0.01)
<u>Overall welfare</u>						
Expenditure (JD/month)	-2.2 (32.6)	+30.9 (39.5)	+15.6 (40.8)	+5.3 (32.8)	<b>+76.8*</b> (39.0)	-4.0 (47.1)
Net income	-22.2 (29.2)	-5.7 (27.1)	-8.8 (33.7)	<b>-66.6*</b> (38.7)	-42.1 (35.8)	<b>-131***</b> (46.6)
Assets	+0.03 (0.03)	+0.04 (0.04)	+0.04 (0.04)	+0.04 (0.03)	+0.02 (0.04)	-0.03 (0.04)
Sample size for comparison	1,914	1,443	1,389	2,359	1,559	1,418

Notes: All estimates are difference-in-differences estimates for coefficient  $\kappa_t$  for period  $t=1$  (after the intervention). Standard errors are shown in parentheses. Statistical significance is denoted as follows: \*\*\*  $p<0.01$ ; \*\*  $p<0.05$ ; \*  $p<0.1$ . The specification controls for time and household fixed effects, and includes additional time-varying controls for the number of refugees arriving in the sample area (a demand shock) and a household-specific wealth index yield very similar results (for alternative results omitting the controls and fixed effects, see Appendix 3). The subsample comparisons are as follows: WNP – Water Network Project treatment zones and matched control zones; WWNP – Wastewater Network Project treatment zones and matched control zones; Both – Water and Wastewater Network Project treatment zones and matched control zones.

<sup>1</sup> Measured on a four point scale (1 = excellent; 4 = poor)

<sup>2</sup> Water samples were only collected and analyzed in Zarqa

<sup>3</sup> The regressions for this outcome do not control for the time-varying factors because we use the full utility database, rather than restricting to the survey sample.

**Table 4.** Summary of main impacts on small enterprise behaviors and outcomes

Outcome	DiD impact of intervention – relative to non-intervention areas in Zarqa, by subsample			DiD impact of intervention – relative to non-intervention areas in Amman, by subsample		
	(1) WNP	(2) WWNP	(3) Both	(4) WNP	(5) WWNP	(6) Both
<u>Water supply</u>						
Piped water is primary source	-0.05 (0.10)	n.a.	-0.01 (0.10)	-0.07 (0.11)	n.a.	+0.09 (0.12)
Hours piped water, for days w/water	-1.8 (2.5)	n.a.	0.25 (2.5)	<b>-8.1***</b> (3.0)	n.a.	-3.7 (3.0)
Water consumption (m <sup>3</sup> /month)	+16.9 (16.8)	n.a.	+24.7 (19.2)	+12.5 (16.5)	n.a.	+25.9 (21.1)
Reported water interruption	+0.03 (0.11)	n.a.	-0.03 (0.12)	+0.03 (0.14)	n.a.	+0.05 (0.15)
Expenditure on water from vendors (JD/month)	+0.21 (0.46)	n.a.	-0.35 (0.49)	-0.54 (0.54)	n.a.	<b>-1.50**</b> (0.61)
Expenditure on water, all sources (arcsin, JD/month)	+0.22 (0.28)	n.a.	-0.02 (0.31)	<b>-0.57*</b> (0.33)	n.a.	-0.01 (0.34)
<u>Wastewater management</u>						
Use of some wastewater system	n.a.	-0.12 (0.08)	+0.02 (0.08)	n.a.	-0.16 (0.10)	+0.05 (0.08)
Sewer connection	n.a.	<b>-0.18**</b> (0.09)	-0.09 (0.09)	n.a.	-0.11 (0.10)	+0.02 (0.09)
Cost of wastewater management (arcsin, JD/month)	n.a.	-0.22 (0.32)	<b>-0.46*</b> (0.26)	n.a.	+0.05 (0.41)	-0.04 (0.37)
<u>Overall welfare</u>						
Expenditure (arcsin, JD/month)	<b>-0.46***</b> (0.15)	-0.02 (0.15)	<b>+0.36*</b> (0.18)	<b>-0.34**</b> (0.17)	+0.07 (0.18)	<b>-0.39*</b> (0.21)
Asset value (arcsin, JD)	+0.25 (0.33)	+0.09 (0.36)	<b>-0.77***</b> (0.28)	+0.24 (0.30)	-0.16 (0.31)	-0.31 (0.31)
Land value (arcsin, JD)	+0.57 (0.35)	+0.28 (0.45)	-0.50 (0.37)	+0.60 (0.43)	+0.41 (0.53)	-0.50 (0.34)
Sample size for comparison	246	229	216	156	139	239

Notes: All estimates are difference-in-differences estimates for coefficient  $\kappa_t$  for period  $t=1$  (after the intervention). Standard errors are shown in parentheses. Statistical significance is denoted as follows: \*\*\*  $p<0.01$ ; \*\*  $p<0.05$ ; \*  $p<0.1$ . The specification controls for time and enterprise fixed effects, as well as the following other time-varying factors: reported complaints about sewer overflows; respondent years with enterprise, number of total employees, reported obstacles to growth, and frequency of water interruptions. Alternative specifications without controls yield very similar results (see Appendix C). The subsample comparisons are as follows: WNP – Water Network Project treatment zones and matched control zones; WWNP – Wastewater Network Project treatment zones and matched control zones; Both – Water and Wastewater Network Project treatment zones and matched control zones.

**Table 5.** Summary of main impacts on farm behaviors and outcomes

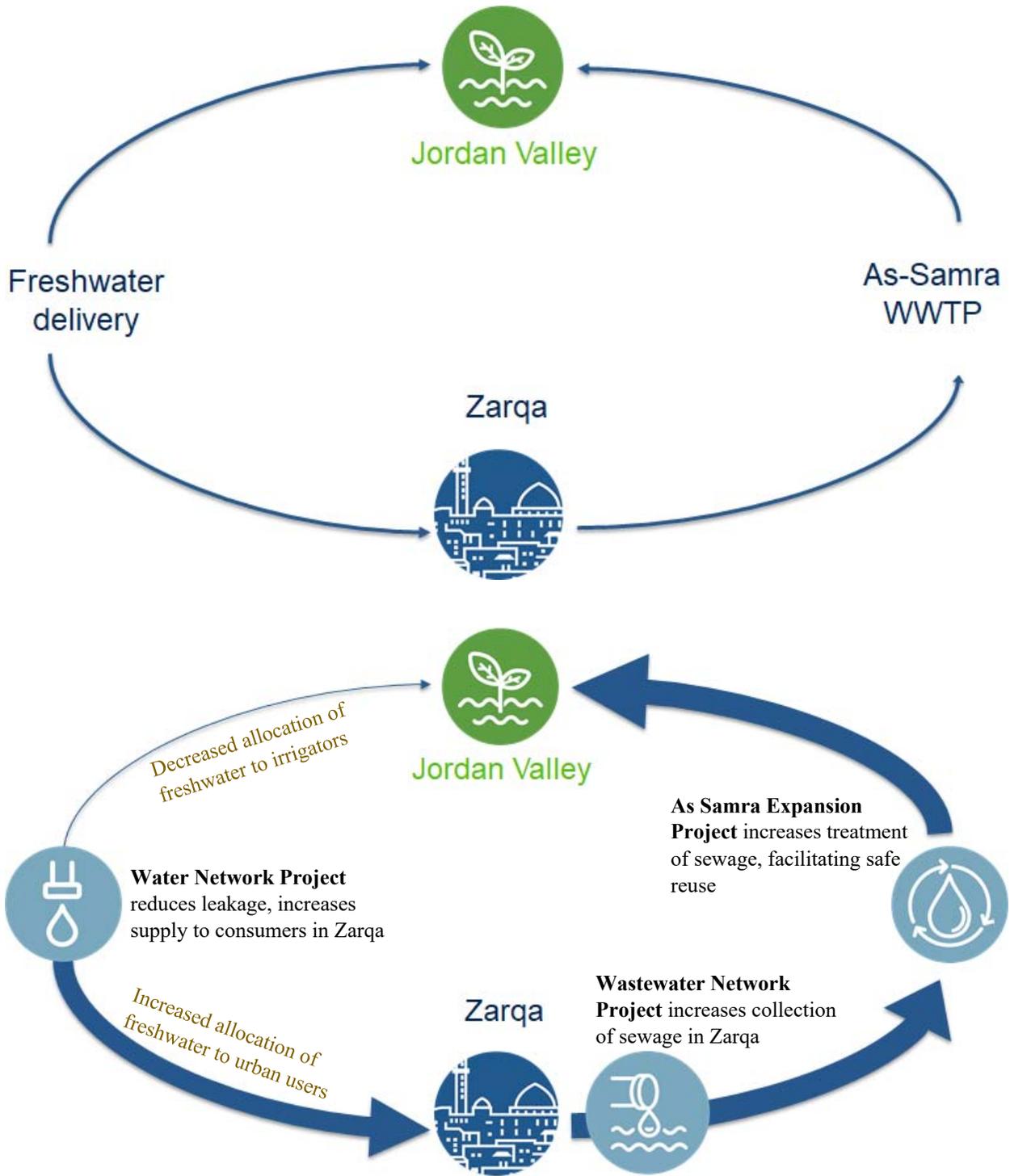
Outcome	DiD impact of intervention – relative to non-intervention areas in Zarqa, by subsample				
	(1) JV1	(2) JV2 (Treatment)	(3) JV3 (Treatment)	(4) JV4	(5) Highlands (Treatment)
Wastewater use in irrigation	<b>+0.10**</b> (0.04)	<b>+0.14***</b> (0.04)	<b>+0.15***</b> (0.04)	<b>-0.47**</b> (0.04)	<b>+0.08*</b> (0.04)
Perceived water quality <sup>1</sup>	<b>-0.81***</b> (0.3)	<b>-0.53*</b> (0.31)	-0.14 (0.31)	<b>+1.40***</b> (0.31)	0.03 (0.34)
Irrigated area (dunum)	<b>+6.0*</b> (3.4)	<b>+8.4**</b> (3.5)	<b>-9.0**</b> (3.6)	<b>-19.9***</b> (3.5)	<b>+12.5***</b> (3.8)
Farm revenue (JD/yr)	<b>+91823*</b> (48757)	-60459 (51173)	-58896 (51336)	<b>-91772*</b> (51908)	<b>+186422***</b> (55180)
Farm profit (JD/yr)	+67362 (47385)	-64974 (49696.95)	-46449 (49854)	-50554 (50473)	<b>+153102***</b> (53656)
Farm land value (JD)	<b>+62124***</b> (18012)	+25717 (19918)	<b>-61272***</b> (19275)	-6484 (21277)	<b>-42200*</b> (22611)

Notes: All estimates are difference-in-differences estimates for coefficient  $\kappa_t$  for period  $t=1$  (after the intervention). Standard errors are shown in parentheses. Statistical significance is denoted as follows: \*\*\*  $p<0.01$ ; \*\*  $p<0.05$ ; \*  $p<0.1$ . The specification controls for time and farm fixed effects. The subsamples are as follows: JV1 is furthest north in the Jordan Valley, and represents a set of farms that were mostly unaffected by the Compact since their water supply is independent of the Zarqa system; JV2 and JV3 represent areas where flows of recycled wastewater newly arrived (JV2) and increased substantially (JV3); JV4 represents an area that already had substantial flows of recycled water prior to the investment; Highlands farms, finally, are located along the Zarqa River and also received access to more steady water supply.

<sup>1</sup> Measured on a ten point scale (1 = poor; 10 = excellent)

**Table 6.** Summary of utility performance indicators, relative to other urban utilities in Jordan

Result	Indicator	Utility	2008	2009	2010	2011	2012	2013	2014	2015	2016	2017	2018	Data Source
Reduced NRW	Indicator: Total losses													
	Total losses; network (%)	Aqaba	20.6	23.9	21.1	22.2	25.9	25.8	27.8	27.7	24.8	25.4	n.d.	MWI, PMU, Utilities
		Amman	40.1	38.3	37.6	32.3	41.2	40.5	47.3	46.2	46.5	45.7	n.d.	
		Zarqa	56.1	54.9	54.9	56.9	55.0	56.2	65.3	63.3	60.6	58.9	58.3	
	Total losses; network (L / Subscriber / Day)	Aqaba	445	484	434	421	488	461	504	475	407	386	n.d.	MWI, PMU, Utilities
		Amman	340	308	302	243	308	321	446	404	403	398	n.d.	
		Zarqa	583	563	555	571	502	550	788	732	663	636	621	
	Indicator: Pipe breaks/bursts per km of mainlines													
	Main bursts/ 100km	Aqaba	139	100	71	76	72	73	90	70	59	n.d.	n.d.	MWI, PMU, Utilities
		Amman	109	89	62	57	45	61	73	64	68	63	n.d.	
Zarqa		n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	175	144	96	88	91		
Service leaks / 1000 connections	Aqaba	122	121	82	192	153	119	116	124	98	n.d.	n.d.	MWI, PMU, Utilities	
	Amman	221	200	172	158	118	179	175	160	140	137	n.d.		
	Zarqa	n.d.	n.d.	n.d.	n.d.	n.d.	n.d.	158	239	155	161	184		
Increased revenue to utility	Utility revenue (2015 JD/m <sup>3</sup> sold)	Aqaba	1.08	0.91	0.90	0.93	1.02	1.04	1.03	1.08	1.07	1.04	n.d.	MWI, PMU, Utilities
		Amman	1.40	1.17	1.19	1.19	1.35	1.40	1.49	1.29	1.34	1.28	n.d.	
		Zarqa	0.90	0.70	0.61	0.62	0.71	0.60	0.53	0.81	0.97	0.88	0.87	
Increased cost recovery by utility	Billing efficiency (%)	Aqaba	91.5	91.5	91.5	91.5	91.5	91.5	91.5	91.3	92.4	91.8	n.d.	MWI, PMU, Utilities
		Amman	97.0	97.0	97.0	97.0	68.0	99.3	99.3	99.2	99.5	99.0	n.d.	
		Zarqa	n.d.	n.d.	n.d.	n.d.	90.0	90.0	90.0	80.8	90.4	98.5	98.9	
	Collection efficiency (%)	Aqaba	99.4	101.0	97.9	92.7	95.3	94.5	92.9	97.9	99.2	96.0	n.d.	MWI, PMU, Utilities
		Amman	96.0	96.0	96.0	96.0	96.0	100.1	97.4	95.8	95.1	97.8	n.d.	
		Zarqa	n.d.	n.d.	n.d.	64.6	108.4	72.9	85.1	92.1	103.1	96.0	91.9	
	Operating Cost Recovery Ratio (OCRR)	Aqaba	1.43	1.26	1.32	1.34	1.34	1.36	1.28	1.36	1.32	1.24	n.d.	MWI, PMU, Utilities
		Amman	1.07	1.09	1.10	1.07	1.03	1.10	1.00	1.09	1.07	1.15	n.d.	
		Zarqa	0.87	0.85	0.70	0.74	0.84	0.70	0.59	0.88	0.83	0.72	0.68	



**Figure 1.** Qualitative depiction of (Top) pre- and (Bottom) expected post- Jordan Compact situations

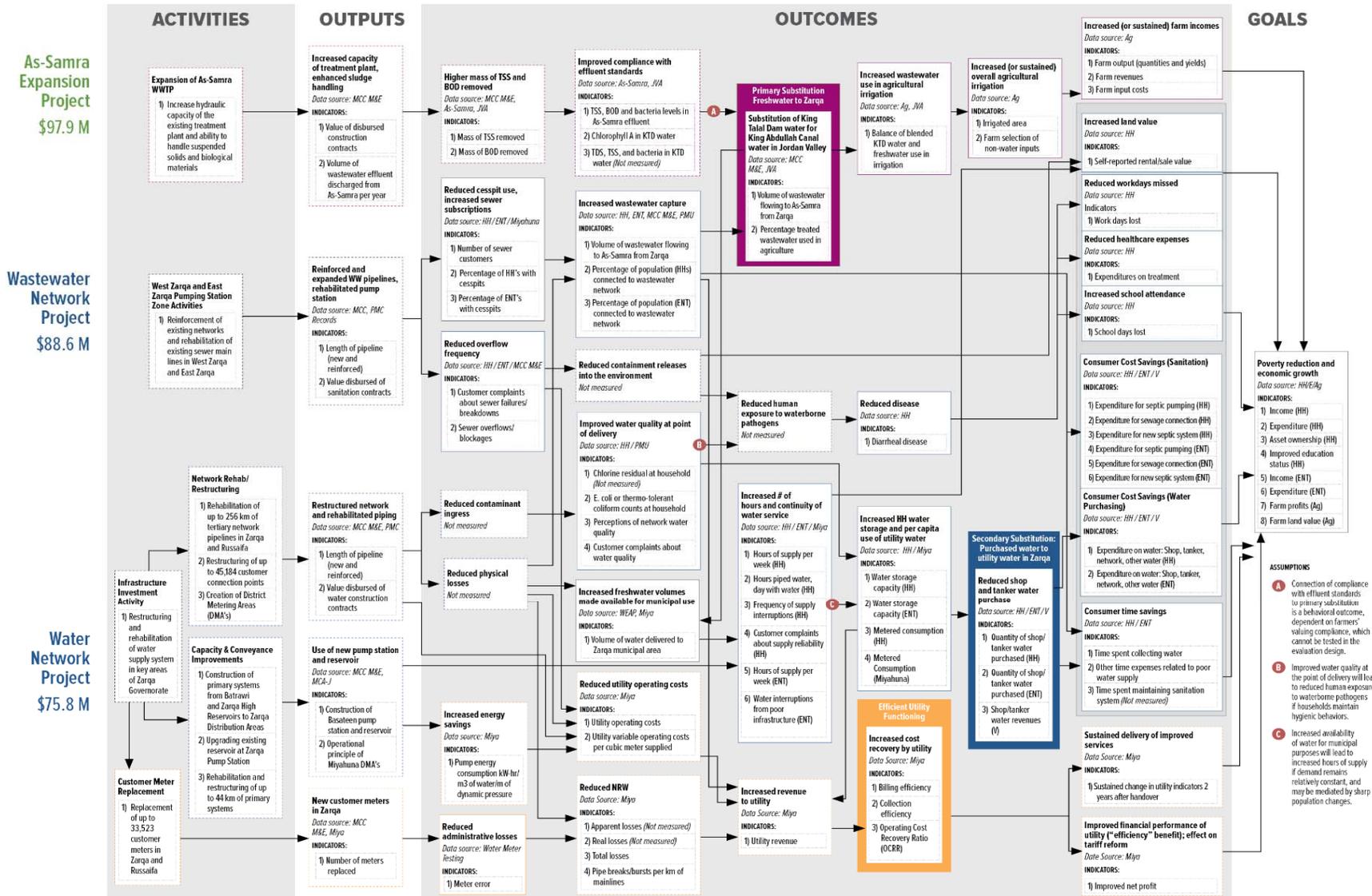
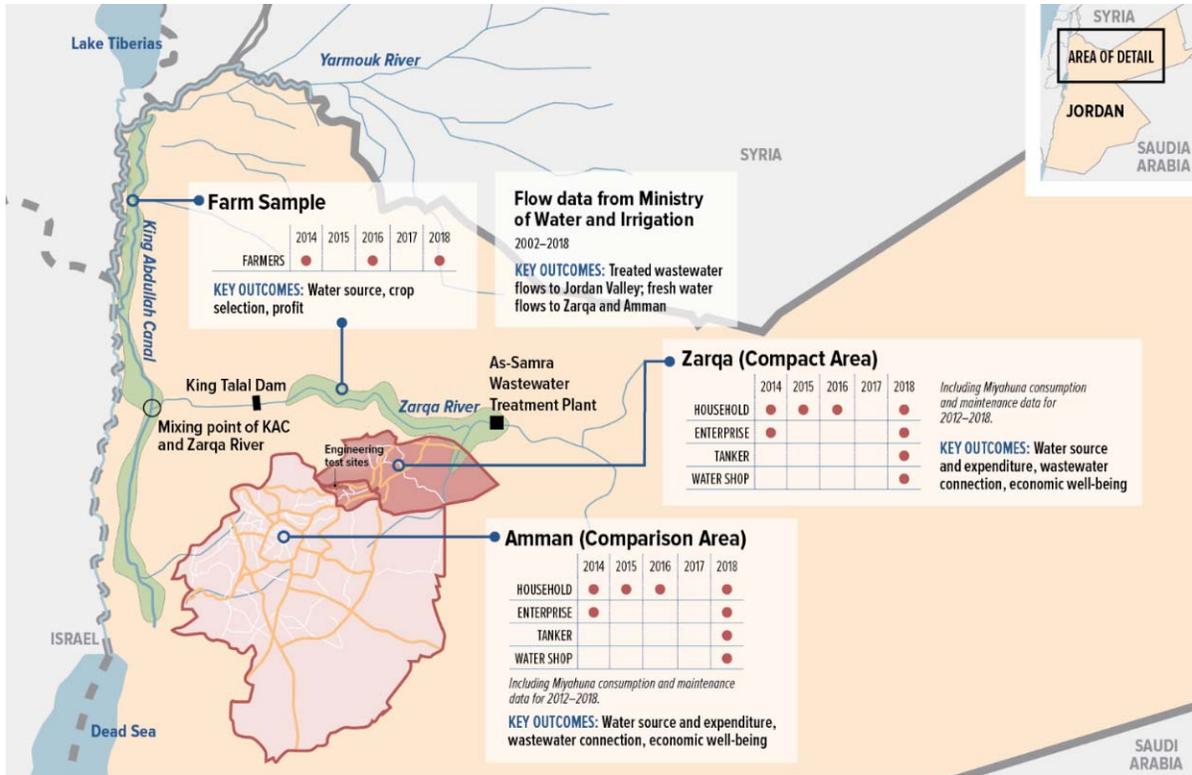
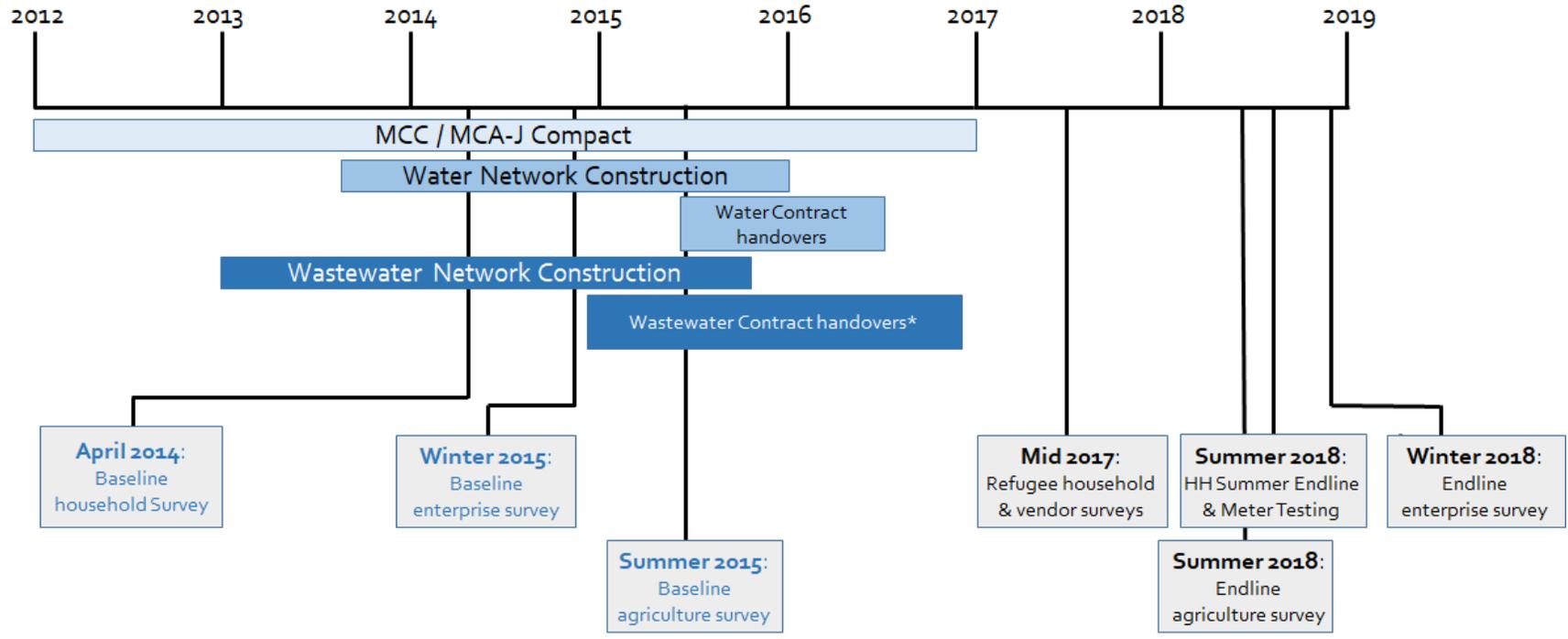


Figure 2. Full program theory of change, as elicited through participatory stakeholder consultations



**Figure 3.** Locations and timing of data collection activities to support the evaluation



**Figure 4.** Timing of data collection relative to infrastructure intervention [Note that baseline surveys were conducted prior to any infrastructure operations, except for a few small wastewater Contract handovers preceding the baseline agriculture survey]