

WORKING PAPER NO. 2012-01

**COST-EFFECTIVE CONSERVATION PLANNING:
TWENTY LESSONS FROM ECONOMICS**

By

Joshua M. Duke., Steven J. Dundas. and Kent D. Messer.

WORKING PAPER SERIES



Alfred Lerner College
of Business & Economics

DEPARTMENT OF ECONOMICS

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Joshua M. Duke
University of Delaware
Department of Food and Resource Economics
213 Townsend Hall
Newark, DE 19716
United States of America

Steven J. Dundas
North Carolina State University
Department of Economics
4129 Nelson Hall
Raleigh, NC 27695
United States of America

Kent D. Messer
University of Delaware
Department of Food and Resource Economics
213 Townsend Hall
Newark, DE 19716
United States of America

Running Head: COST-EFFECTIVE CONSERVATION PLANNING

Keywords: Conservation planning, cost-effectiveness, nonmarket valuation, benefit-cost targeting, optimization, prioritization

JEL Codes: Q18, Q24, Q57, Q58.

Corresponding Author

Joshua M. Duke
Department of Food and Resource Economics
531 South College Avenue
Newark, DE 19716
United States of America
302-831-2511
duke@udel.edu

1 **Abstract**

2 Economists advocate that the billions of public dollars spent on conservation should be allocated
3 to achieve the largest possible social benefit. This is what we term “cost-effective conservation”--
4 a process that incorporates both benefits and costs that are measured with money. This
5 controversial proposition has been poorly understood and not implemented by conservation
6 planners. Drawing from evidence from the largest conservation programs in the United States,
7 this paper seeks to improve the communication between economists and planners and overcome
8 resistance to cost-effective conservation by addressing the open questions that likely drive
9 skepticism among non-economists and by identifying best practices for project selection. We first
10 delineate project-selection strategies and compare them to optimization. Then we synthesize the
11 body of established research findings from economics into 20 practical lessons. Based on theory,
12 policy considerations, and empirical evidence, these lessons illustrate the potential gains from
13 improving practices related to cost-effective selection and also address how to overcome
14 landowner-incentive challenges that face programs.

15 **1. Introduction and Policy Setting**

16 Governments should use conservation policies to enhance the benefits to society in lieu of fully
17 functional markets for ecosystem services. These policies conserve land by requiring or
18 incentivizing landowners to protect habitat for endangered species, control erosion, enhance
19 riparian buffers and wetlands. They also preserve agricultural and forest land by purchasing land
20 outright or purchasing conservation easements to preclude development. While conservation
21 activity exists throughout the world, most of these efforts are less effective than they could be.
22 Drawing from evidence from conservation programs in the United States this paper reviews the
23 process by which governments and large non-governmental organizations pursue conservation
24 and recommends best practices that will enhance conservation outcomes.

25 At a fundamental level, economists recommend that conservation planning should
26 account for all of the social benefits resulting from a project, regardless of to whom they accrue,
27 rather than focusing on environmental benefits alone. These policies should ensure that these
28 social benefits are as large as possible given constrained conservation budgets. *Cost-effective*
29 project selection is a process that incorporates both benefits and costs that are measured
30 commensurately with money and seeks to maximize the conservation outcomes important to the
31 public. This type of approach delivers the “best bang for the buck” and any other selection
32 approach sacrifices some achievable benefits. While an *economically efficient* solution is to
33 pursue all conservation projects for which the social benefit exceeds the social cost,
34 unfortunately, limited budgets for conservation generally preclude such an effort. Thus, we focus
35 on cost-effectiveness rather than efficiency and study the complexities of optimal project

36 selection. These complexities include conflicting incentives, selection challenges, dynamic
37 effects, interdependencies, and uncertainties.

38 The use of the terms *cost-effective conservation* in this review should not be confused
39 with *cost-effectiveness analysis*, a decision science method, which is common in health
40 economics and has been used in some literature related to conservation selection. Cost-
41 effectiveness analysis explicitly excludes measuring benefits in monetary terms, which we show
42 in this manuscript can often lead to suboptimal conservation outcomes.

43 Allocating funds to achieve the greatest possible conservation benefit—the economic
44 concept of cost-effectiveness—remains controversial among academics and lacks widespread
45 adoption by conservation planners, policymakers, conservation program architects, and funders
46 (hereafter referred to collectively as “planners”). Although many papers in the conservation
47 planning literature identify the advantages of cost-effective conservation, several recent papers
48 have argued against this growing push because the complex interaction between humans and
49 nature exceeds the capacity of traditional economic methods (Arponen et al. 2010; Gowdy et al.
50 2010). Such critiques arise close to the heart of economics and complement long-standing
51 objections to the use of benefit-cost analysis. For instance, Odling-Smee (2005:616) points out
52 that some see efforts to monetize nature as violating “ethical and spiritual dimensions of
53 conservation.” While acknowledging these critiques, we believe that modern economic
54 valuation techniques can provide some measurement of these values and targets this manuscript
55 at the practical problems of improving the effectiveness of current conservation programs.

56 Conservation expenditures are rapidly increasing. The U.S. Farm Bill covering 2008-
57 2012 allocates \$11.7 billion to working lands programs such as the Environmental Quality

58 Incentives Program (EQIP), \$1 billion to agricultural land preservation, and \$13 billion to land
59 retirement programs such as the Conservation Reserve Program (CRP) (author calculation based
60 on data reported in Claassen (2010)). U.S. federal conservation expenditures represent a \$7.8
61 billion increase over the prior baseline (Hajkowicz et al. 2009), and yet this still understates
62 conservation efforts because it does not include state, local, and nongovernmental conservation
63 activity. Private U.S. land preservation by 1,667 land trusts and nongovernmental organizations
64 had protected 37 million acres by 2005, with total preservation doubling between 2000 and 2005
65 (Aldrich & Wyerman 2006). Furthermore, the federal government and states spent at least \$11.1
66 billion on endangered species recovery between 1989 and 2004 (Langpap & Kerkvilet 2010).
67 Conservation efforts in the European Union (EU) may exceed those in the U.S.; for instance,
68 between 2007 and 2013 the EU plans to spend €35.4 billion on agri-environmental payments
69 alone (author calculation based on data from the EU (2009)). Governments throughout the world
70 pursue conservation. For instance, in New South Wales, Australia, the Environmental Services
71 Scheme provides incentives to alter private land management in an effort to improve delivery of
72 environmental services (Oliver et al. 2005). Finally, China’s Sloping Land Conversion Program,
73 perhaps the world’s largest conservation program with an estimated budget of \$48 billion, seeks
74 to convert crop and wasteland to forests (Xu et al. 2010).

75 Evidence suggests challenges in communication between planners, policymakers, and
76 economists. Banzhaf (2010: 592), in part, faults economists’ for their “lack of interest in making
77 academic work accessible”. Prendergast et al. (1999: 484) cites a lack of awareness and
78 understanding as possible obstacles to using theoretically driven conservation planning, as well
79 as limited funds and even “antipathy” toward “prescriptive” selection tools. Planners may also

80 resist cost-effectiveness because they are not familiar with optimization mathematics and lack
81 tools for implementation amongst numerous other reasons (Ferraro and Pattanayak 2006; Messer
82 et al. 2011). Calls for greater dialogue and collaboration are long-standing (Prendergast et al.
83 1999; Armsworth et al. 2004). It is this lack of constructive communication, cooperation, and
84 resistance to economic approaches that motivates this synthesis.

85

86 **2. Methods**

87 The scientific literature on the practice of cost-effective conservation is vast, and a book-length
88 treatment would be required to review it all. In addition, there is an applied literature that
89 evaluates certain programs and a call for more work in this area (Laycock et al. 2009; Ferraro
90 and Pattanayak 2006). Existing syntheses, therefore, focus on somewhat narrow aspects. One
91 rationale for this work is to present cost-effective conservation in a new and, hopefully, more
92 useful package for planners. This section explains how literature was selected and organized. We
93 briefly review existing approaches before turning to the one in this paper.

94 Claassen et al. (2008) offered a comprehensive review of the CRP and EQIP and found,
95 in part, that existing rules delivered were better than some alternative selection processes, but
96 were still not truly cost effectiveness. Wu (2004) summarized many of the challenges to cost-
97 effective conservation and focused on impediments associated with the policy process and
98 complexities associated with the resources targeted for protection. Newburn et al. (2005)
99 comprehensively assesses cost-effective conservation in light of vulnerability. Sarkar et al.
100 (2006) synthesized the concepts, techniques, and software available for optimal biodiversity
101 conservation planning. Most similar in approach to our paper is Wilson et al. (2009), which

102 offered lessons about setting priorities in biodiversity planning. Wilson et al. (2009) identified
103 specific challenges to prioritizing conservation—including temporal issues, uncertainty, and
104 spatial heterogeneity, and drew conclusions about the need for location-specific planning.

105 Unlike prior syntheses, we offer 20 lessons to assist planners make more cost-effective
106 decisions with their limited resources, thereby increasing the supply of ecosystem services.
107 Practical guidance grounded in research is needed because, as Prendergast et al. (1999) argued,
108 the benefits of cost-effectiveness frequently fail to reach planners who make actual conservation
109 decisions. Several lessons presented in this paper arise from recent research while others are
110 practical guidance original to this work. In addition, this paper offers a broad, and therefore
111 shallow, perspective to complement other syntheses offering topical depth. Finally, the paper
112 also highlights areas where research has identified significant challenges in conservation
113 planning. Explicit recognition of the current challenges facing cost-effective conservation
114 hopefully will help build credibility with potential adopters and clarify future research agendas.

115 Economic research in conservation tends to focus on empirical analyses of and challenges
116 to the practice of conservation because the theory of optimal selection is relatively
117 straightforward. Therefore, the next section briefly summarizes the theory and defines cost-
118 effective conservation. We then distill the literature into 20 best-practice lessons and organize
119 these lessons into five sections (summarized in table 1): optimal selection, benefits, costs,
120 budgets, and incentive problems.

121

122 **3. Theory: Cost-Effective Project Selection**

123 Planners typically pursue *conservation benefits*, such as biodiversity, habitat provision,
124 agricultural land quality, and air quality, and use *benefit indices* to measure the benefits that
125 would arise from investment in a project. For example, the CRP and the Wetlands Reserve
126 Program in the United States assign relative weights, which are periodically adjusted for each
127 type of environmental benefit targeted (Cattaneo et al. 2006). These weights substantively impact
128 project priorities but there is little guidance on how to sum these benefits when they are
129 incommensurate. Hajkowitz et al. (2009) conducted an assessment of programs that use benefit
130 indices and recommended better incorporation of social preferences in the weights (measured
131 with appropriate techniques) and development of standardized indices.

132 Measuring the costs of conservation, such as acquisition, transaction, monitoring, and
133 stewardship costs, is more straightforward because existing markets often reveal these values.
134 Nevertheless, Ando et al. (1998) notes that costs are not widely incorporated in conservation
135 decisions. Ignoring costs may have once made sense when the goal was protection of unique
136 natural amenities such as the national parks of Yellowstone or the Grand Canyon. However,
137 current conservation practices extend to many settings where programs must decide where to
138 invest their limited funds among a number of high-quality projects that are close substitutes in
139 terms of environmental benefits but differ substantially in cost. In these settings, paying too
140 much can significantly reduce the benefits from conservation efforts.

141 Selection strategies that focus on only one measure—benefit targeting or cost targeting—
142 consistently lead to suboptimal results. Strategies that include both costs and benefits, such as
143 benefit-cost targeting, benefit maximization targeting, and mathematical programming methods,
144 are being adopted, albeit slowly. This section distinguishes these techniques.

145 *Benefit targeting (BT)*, also termed “benefit ranking” or “rank-based model” ranks
146 projects according to their environmental benefit and selects the highest-ranking ones until the
147 budget is exhausted (Ferraro 2003). It is used frequently for private and public conservation
148 programs, such as the U.S. Fish and Wildlife Service (Wu 2004), for the establishment of
149 national parks (Babcock et al. 1997; Wu et al. 2001). BT has intuitive appeal to many
150 conservationists, who are drawn to projects with the largest environmental benefits. However,
151 BT ignores cost as a selection criterion, and the outcome is likely to be cost-ineffective because
152 the budget can be exhausted by a couple of high-benefit, high-cost projects (Messer 2006).

153 *Cost targeting (CT)* ranks projects solely by acquisition cost and selects the least
154 expensive ones until the budget is depleted—a “bargain shopper” tactic (Ferraro 2003). In
155 practice, CT tends to maximize acreage rather than net benefit (Babcock et al. 1997). Pure CT
156 seems to be relatively rare in practice, though examples exist. Babcock et al. (1997), for
157 example, framed the CRP’s early efforts as equivalent to CT. Another related example is the
158 Delaware Agricultural Lands Preservation (DALP) program that uses a reverse auction—an
159 auction with one buyer and multiple sellers—and selects projects based on the level of discount
160 offered by owners on the appraised development increment (Messer and Allen 2010).

161 *Benefit targeting with a cost adjustment* is similar to BT but scores conservation costs as
162 a nonmonetary benefit measure. For example, Ribaud et al. (2001) calculated that the cost
163 factor score used by the CRP represents 27% of total possible points, subject to soil quality, in
164 the Environmental Benefits Index. While this strategy may have intuitive appeal because it
165 seems to analyze costs and benefits jointly, it is not truly cost-effective (Hajkowicz et al. 2009)

166 as it is easy to construct examples where scoring costs as a benefit leads to sub-optimal
167 environmental results.

168 *Benefit-cost targeting (BCT)* selects projects with the highest benefit-cost ratios until the
169 budget is exhausted. This approach ensures selection of individual projects that have the highest
170 benefit per dollar, which will achieve no worse and typically greater cost-effectiveness than BT
171 or CT (Babcock et al. 1996). This characteristic leads many economists to promote BCT (Ferraro
172 2003). In fact, U.S. federal programs, such as the CRP and EQIP, use a version of BCT that
173 seeks to maximize environmental benefit per dollar spent (Wu et al. 2001), however, since cost is
174 measured as a benefit index true cost-effectiveness is not achieved.

175 Wu et al. (2001) and Wu (2004) described how characteristics of commodity markets
176 might create secondary impacts that prevent BCT from maximizing total net social benefits in
177 some conservation settings. These technical distinctions led to an improved selection strategy:
178 benefit-maximization targeting. *Benefit-maximization targeting* selects projects to minimize
179 increases in commodity output prices and, thus, slippage (described later) and achieves the same
180 level of environmental benefit as BCT but at a lower cost (Wu 2004). In principle, benefit-
181 maximization targeting is fully cost-effective; however, the literature has tended to employ
182 relatively simple problems to demonstrate this technique. Because project selection occurs in a
183 complex world of constraints and interdependencies, true cost-effectiveness requires even more
184 advanced techniques.

185 *Optimization* involves a set of mathematical programming algorithms, such as binary
186 linear programming and goal programming, from operations research that seek to maximize total
187 net benefits and achieves cost-effectiveness in more complex situations, such as a need to enroll

188 a minimum number of acres, to maximize the number of species preserved, to select a minimum
189 number of projects from a particular region, or to meet disparate goals (Underhill 1994; Sarkar et
190 al. 2006; Balmford et al. 2000; Kaiser & Messer 2011; Fooks & Messer, *forthcoming*).
191 Optimization algorithms can identify optimal selections when ecological complexities such as
192 thresholds introduce jointness to the selection of projects, a problem investigated by Wu et al.
193 (2000) and Wu (2004). In addition, these techniques can offer slight advantages over iterative
194 selection techniques, such as BCT, by adjusting to account for budget remainders (Messer 2006).

195

196 **4. Twenty Lessons for Cost-Effective Selection Processes**

197 **4.1 Optimal Selection**

198 **Lesson 1: Benefit targeting and cost targeting can lead to suboptimal project selection.** The
199 weakness of these approaches can be demonstrated with a numerical example provided in table
200 2, which gives hypothetical data for prioritization of six conservation projects using costs and
201 monetized benefits. The second panel of table 2 compares the projects selected with a budget of
202 \$6 by several ordinal (ranking) and cardinal (quantity) prioritizations arising from BT (column I)
203 and CT (column J) with the selections made by optimization using monetized benefit-cost ratios
204 (column L). In this example, net benefits are maximized at \$44 by selecting projects A, B, and C.
205 BT and CT prioritizations are suboptimal at a net benefit of \$40 and \$43 respectively.

206 Empirical evidence supports the hypothetical example, and the magnitude of the cost-
207 ineffectiveness can be substantial. In an application to endangered species protection, Ando et al.
208 (1998) found savings of as much as 75% when costs were systematically accounted for. Messer
209 and Allen (2010) examined the DALP program and showed that optimal selection would have

210 preserved the same number of acres with an equal benefit score but would have saved
211 approximately \$21 million relative to DALP's CT system (more than 20% savings) and
212 substantially more if DALP had used BT. In the case of conservation of terrestrial vertebrates in
213 Oregon, incorporating land costs would have generated a ten-fold improvement in cost-
214 effectiveness (Polasky et al. 2001). Recent adoption of BCT in Baltimore County, Maryland,
215 resulted in protection of an additional 680 high-quality agricultural acres—saving \$5.4 million—
216 compared to BT in just three years (Kaiser & Messer 2011:271).

217 Fully optimal methods require substantial data. However, several studies suggest that
218 policymakers might approach optimal selection even if some data are unavailable. This depends
219 on what one knows about the distribution of unobserved costs and benefits. When benefits and
220 costs are uncorrelated, BT performs better when benefits vary more than costs—and vice versa
221 for CT (Babcock et al. 1997). A number of studies have examined optimal selection with
222 observed data on variability of costs and/or benefits (Ando et al. 1998; Balmford et al. 2003;
223 Ferraro 2003; Perhans et al. 2008) and evaluated selection performance without complete data
224 (Babcock et al. 1997; Ferraro 2003; Perhans et al. 2008). In general, positive statistical
225 correlation between a project's costs and benefits tends to improve the performance of BCT
226 relative to BT or CT, while a negative correlation leads to more similar performances for the
227 three methods (Babcock et al. 1997).

228 **Lesson 2: Efforts to distribute conservation funds evenly across political**
229 **jurisdictions will tend to be suboptimal.** The political process and perceptions of fairness may
230 introduce constraints. For example, the CRP limits program participation to 25% of cropland in
231 any county to protect local economies (Sullivan et al. 2004), and Pennsylvania's agricultural land

232 preservation program distributes money to all participating counties, each administering
233 individual programs (3 P.S. § 914.1(b,h)). Such constraints reduce cost-effectiveness because
234 they restrict the feasible set of solutions and, by definition, cannot improve the cost-effectiveness
235 of the solution (Kaiser & Messer 2011). These constraints also can work against efforts to target
236 conservation in settings where biological thresholds are important (Wu et al. 2000, Wu &
237 Boggess 1999; Wu & Skelton-Groth 2002; Wu 2004). The political reality, however, is that
238 distributing funds across jurisdictions may help secure broad legislative support for a program.
239 Likewise, nongovernmental organizations may win political favors or improve fundraising by, at
240 times, focusing on high-profile projects.

241

242 **4.2 Benefits**

243 **Lesson 3: Measure conservation benefits that are positive externalities.** Gardner (1977)
244 provided an early summary of fundamental economic concerns about emerging land preservation
245 policies. Because some of its points remain underappreciated while others have been
246 misunderstood, revisiting Gardner’s arguments is worthwhile.

247 Gardner notes that policy interventions in land markets can increase total social benefits
248 if there is a market failure, but they reduce the productivity of scarce resources if no failure
249 exists. Gardner found a land market failure in the under-provision of public goods—in other
250 words, land markets provide too few ecosystem services. Termed *external benefits* or *positive*
251 *externalities*, such services include wildlife habitat, water quality protection, scenic views, and
252 carbon sequestration. Landowners rationally undersupply them because existing markets do not
253 fully capture the social benefits of their decisions. Gardner’s argument implies that external

254 benefits should be measured and then policy should internalize them by incentivizing
255 conservation. Gardner correctly anticipated that policymakers would incentivize easy-to-measure
256 benefits such as soil quality and, thus, cautioned that increasing the supply of such benefits does
257 not clearly enhance resource allocation efficiency because no obvious market failure exists for
258 soil quality (i.e., farmers already pay more for high-quality land). Instead, Gardner argued that
259 appropriate conservation benefit measures reflect factors that are external to markets and are
260 associated with benefits that accrue to neighbors and the general public.

261 **Lesson 4: Measure benefits to the public, not to experts.** The logic for this potentially
262 controversial lesson is that the public is the group that receives the services. The economic
263 literature offers evidence that the conservation preferences of experts may or may not diverge
264 from those of the public (Strager & Rosenberger 2006; Columbo 2009). While this lesson may
265 not be relevant to private conservation organizations as they are driven by their donor priorities,
266 it does apply to government agencies and perhaps also to larger conservation organizations.
267 Some public preferences can be measured or estimated (see Kline 2006). We acknowledge that
268 this lesson may be challenging to follow when the conservation benefits are associated with
269 ecosystem services that the public is unlikely to fully understand, such as implications of specific
270 pollutant loads or habitat needs for an endangered species.

271 **Lesson 5: Monetize benefit measures.** Monetized benefit measures (conservation
272 benefits measured in dollar terms) are required for cost-effective policy because they must be
273 balanced with the costs of conservation, which are often largely monetized—Kido & Seidl
274 (2008) apply such techniques to develop optimal protected area entry fees. Conservation
275 programs tend to use benefit indices derived from agri-environmental criteria such as soil

276 quality, crop productivity, soil erosion, water quality, and carbon sequestration (Hajkowicz et al.
277 2009). The CRP, for example, uses the Environmental Benefits Index while some agricultural
278 land preservation programs use the Land Evaluation and Site Assessment (LESA) system. EQIP
279 uses a ratio of value of the benefit index (BI) to the cost to achieve statutorily mandated cost-
280 effectiveness in securing environmental benefits (Cattaneo 2003). These indices capture well the
281 services that landowners supply; however, they do not correspond to the value society places on
282 the supply of such services (Smith 2006).

283 Note that efforts to monetize public welfare can lead to systematic biases if income and
284 net-benefit incidence are correlated and wealth is unequally distributed. This is a well-known
285 challenge to all benefit-cost analyses. Also, some find this assertion controversial if one does not
286 believe that values for ecosystem services can be measured monetarily.

287 Fortunately, monetized benefit measurement has advanced considerably over the past
288 three decades. For instance, many applications measure the benefits of preserved land, and these
289 benefits increase on-parcel and off-parcel human welfare (Bastian et al. 2002). Valuation
290 techniques include revealed preferences (such as hedonic analysis) and stated preferences (such
291 as contingent valuation and choice modeling). Future areas of research in this area include the
292 influence of certain amenities, such as public access, spatial relationships, and different
293 agricultural uses (Bergstrom & Ready 2009).

294 Decision-makers have argued, incorrectly as will be shown, that nonmonetized benefit
295 measures (benefit indices) equally promote cost-effectiveness, particularly if the indices use
296 cardinal measures (the index employs units that reflect more than a ranking). Economists and
297 other environmental researchers have employed sophisticated cardinal techniques for

298 aggregating preferences. Techniques include the analytic hierarchy process (see Ananda &
299 Herath 2009) and the logic scoring of preferences (Allen et al. 2011), which can be used with
300 groups of experts or the general public.

301 **Lesson 6: Benefit indices can lead to suboptimal project selection.** Messer & Allen
302 (2010:45–46) demonstrate how benefit indices, which are often averaged for the conservation
303 project as a whole rather than assigned per acre, can lead to scaling problems. In effect, an
304 averaged benefit index will be biased against large projects.

305 Benefit indices also can map poorly into monetized benefits. This can be demonstrated by
306 revisiting the example in table 2. Assume that monetized benefits are shown to be a linear
307 function of the benefit index: $\$B=BI+7$ (column D). Even with this simple, monotonically
308 increasing relationship of just adding 7 (one can readily imagine a more complex relationships
309 between $\$B$ and BI), this example shows that the BI -cost ratio (column K) produces a smaller
310 total net benefit of \$40 than the optimum of \$44 (column L). This result may be counterintuitive,
311 but it occurs because systematic mismeasurement of the monetized benefit reverses the rank of
312 the selected projects. Although the values shown in table 2 were selected to demonstrate these
313 points, the example demonstrates that an ostensibly reasonable cardinal BI can lead to smaller
314 net benefits even when monetized benefits are a simple transformation.

315 **Lesson 7: Targeting conservation benefits leads to greater cost-effectiveness when**
316 **thresholds are present.** Conservation thresholds complicate optimal selection and exist when an
317 environmental benefit depends on achieving some minimum level of conservation (Wu et al.
318 2000; Wu & Skelton-Groth 2002; Wu 2004). Examples are when a minimum amount of habitat
319 is needed to sustain an endangered species or a critical mass of farmland must remain to sustain a

320 region's viable agricultural industry. Wu & Boggess (1999) offered an assessment on how
321 thresholds complicate optimal selection. Wu et al. (2000) and Wu & Skelton-Groth (2002)
322 extended that work with empirical evidence about how targeting conservation leads to greater
323 cost effectiveness when thresholds exist for fish habitat protection.

324 **Lesson 8: Interrelationships (correlations and interactions) among conservation**
325 **projects are often unobserved.** This is especially true when readily available benefit measures
326 such as soil quality drive the selection process. Studies have examined how targeting
327 conservation leads to optimal selections when projects are interrelated (Wu & Boggess 1999).
328 Interrelationships can take many forms. For instance, preserving habitat on two contiguous
329 parcels will likely deliver greater joint benefits than two discontinuous parcels, all else equal. In
330 other words, spatial scale matters and there can be a spatial agglomeration of benefits. An
331 interrelationship also may exist between two different types of ecosystem services, such as
332 riparian protection that improves the land-based and the aquatic habitat. A number of studies
333 have examined efforts involving agglomeration bonuses to incentivize landowners to coordinate
334 their behavior (see Parkhurst et al. (2002); Parkhurst & Shogren (2007); Drechsler et al. (2010)).

335 Many studies have sought to spatially model environmental benefits (see van der Horst
336 (2007)), however, fewer studies have examined monetized benefits spatially (Bateman et al.
337 2003; Hynes et al. 2010; Campbell et al. 2009). van der Horst (2006, 2007) developed a method
338 for considering multiple benefits in space and calculating effectiveness gains from spatial
339 targeting of two benefits, which is then assessed via an analysis of the Farmland Woodland
340 Premium Scheme in Scotland. Wu (2004) argued that lack of information, rather than a failure to

341 recognize the interrelationships, has led to the current policy environment, which tends to focus
342 on specific resources rather than the more complex ecosystems relationships.

343 **Lesson 9: Optimal selection accounts for development risk.** Conservation decisions
344 typically are made with uncertainty about future benefit supply. Some projects supply benefits
345 even in the absence of conservation, while others risk diminution or destruction. Therefore,
346 researchers promote and many planners desire conservation targeted at the most vulnerable
347 benefits first, though there so far is no consensus on how best to do this. For instance, Messer
348 (2006) argues that development threat can be predicted from observable parcel characteristics
349 (location, soil quality, proximity to highways, etc.) that can in turn give weights to various
350 benefit measures prior to optimization. Because development risk tends to vary directly with
351 cost, Newburn et al. (2005) offered an approach to optimal selection (benefit-loss-cost targeting)
352 that allows risk and costs to be assessed jointly. Costello & Polasky (2004) developed an optimal
353 dynamic selection model that accounts for development risk and found that heuristic selection
354 performs reasonably well when a dynamic problem becomes too large. Nonmarket valuation
355 offers an additional perspective as it directly estimates the marginal benefit of preserving lands at
356 various levels of development risk. Johnston & Duke (2007) estimated higher benefits from
357 preservation of parcels at the highest risk of development.

358 **Lesson 10: The policy process impacts the conservation benefit received.** Empirical
359 evidence demonstrates that the public cares about how and by whom conservation benefits are
360 secured, where the policy process refers to the policy used and administering entity. Many
361 policies exist to deliver conservation services and, furthermore, these services can be delivered
362 by governmental agencies or nongovernmental organizations. These groups preserve land with

363 easements or fee simple ownership, and governments can use zoning/regulatory mechanisms.
364 Water quality, for example, may be enhanced by regulations, incentive programs such as the
365 CRP, government-sponsored relocation of nutrients, tax instruments, or nutrient trading.
366 Johnston & Duke (2007) found, in the case of farmland, that mandatory governmental zoning
367 was viewed by the public negatively compared to a voluntary state easement program that was
368 viewed more favorably and therefore delivered higher monetized benefits. Of course, the costs of
369 these efforts can be different as some studies have shown zoning, while controversial, to be
370 relatively low cost and effective (Ozama and Tertley, 2007).

371 **Lesson 11: Markets will tend to capitalize location-specific benefits.** For example, a
372 house will tend to increase in value if it borders a newly protected preserve or farm (Geoghegan
373 2002; Irwin 2002; Netusil 2005; Geoghegan et al. 2003). Property values will even increase if
374 proximity to a conserved area allows for access to newly supplied services such as nature trails.
375 Although potential capitalization does not invalidate conservation benefits, competitive rental
376 markets can drive renters to indifference (Landsburg 1993:34–37), i.e., owners may increase rent
377 to account for the enhanced environment. This obviously represents a potential equity problem:
378 because capital owners tend to be wealthier than nonowners, thus, capitalization will tend to lead
379 to some efficiency mismeasurement (Duke & Johnston 2011). This is an area for future research
380 as researchers have not yet devised definitive advice on how to integrate capitalization into
381 analyses of public good supply. Also, not all conservation benefits will be location-specific (e.g.,
382 endangered species protection) so capitalization will not complicate all selection problems.

383

384 **4.3 Costs**

385 **Lesson 12: Include and fully account for all costs.** Optimal selection requires data on the
386 projects' costs, and Naidoo et al. (2006) offers a thorough accounting of why and how costs
387 should be used in conservation planning. Although markets do supply some project cost data,
388 such as the cost of acquiring the land or easement, economists note that optimality requires
389 accounting for all costs—and this is directly related to a landowner's willingness to participate in
390 programs (Miller et al. 2011). Frequently ignored factors include in-kind costs such as volunteer
391 labor and external costs such as increased nuisance species. Likewise, costs should be estimated
392 for future management and restoration costs. Naidoo et al. (2006:682) offers a typology of these
393 costs, and Wilson et al. (2009:242) presents an extensive list of costs and associated research
394 studies. Moilanen and Arponen (2011) address more complicated planning situations, such as
395 when priorities must be set though future costs are uncertain.

396 **Lesson 13: Costs should be monetized.** Naidoo et al. (2006) describes efforts to proxy
397 with nonmonetized costs and argues that simple averages ignore spatial heterogeneity while
398 more advanced estimates can sufficiently capture variation. Carwardine et al. (2010) extends this
399 work by assessing how sensitive optimal prioritization is to levels of cost uncertainty.

400 **Lesson 14: Sequential assessment of benefits and then costs tends to be suboptimal.**

401 To understand this potential pitfall, consider again the DALP easement program that uses a
402 LESA benefit index to score all applicant parcels and then selects a subset of parcels that exceed
403 a minimum score for further consideration (3 Del. C. § 9-908(a)(4)). The high-scoring parcels
404 are then sorted by the owners' offered discounts (i.e., cost targeting) (3 Del. C. § 9-914(b)(3)).
405 While this selection method analyzes benefits and costs, the sequential approach cannot

406 guarantee optimality. Consider a hypothetical example where high-benefit project A offers a
407 benefit of 10 and a cost of 9, project B offers a benefit of 9 and a cost of 9, and low-benefit
408 projects C, D, and E each offer a benefit of 7 and a cost of 3. Assume the benefits reflect all
409 relevant conservation data. With a budget of 9, cost-effectiveness will select C, D, and E,
410 conserving three projects for total net benefits of 12. Sequential analysis would immediately
411 eliminate C, D, and E and focus on A and B. If A is chosen, the budget would be exhausted and
412 the net benefit would be just one. Thus, the sequential approach may seem to control the cost of
413 seeking high-benefit projects, but it is generally suboptimal.

414

415 **4.4 Budgets**

416 **Lesson 15: Large budgets allow conservation of all projects, any selection strategy will be**
417 **optimal** (Babcock et al. 1997). While this lesson is straightforward, it is important to recall that
418 the differences in selection strategy arise when budgets are limited. Furthermore, the more
419 limited the program's budget, the greater the potential gain from optimal prioritization.

420 **Lesson 16: Optimization improves cost-effectiveness when budget remainders are**
421 **significant.** Remainders are a significant problem with limited budgets. Large remainders are
422 most likely when budgets are severely limited, especially when project costs are high relative to
423 the budget, when agencies cannot implement projects in fractions, and when budgets cannot be
424 carried over into new periods. Such gains are a key difference between BCT and optimization
425 (Messer 2006). Consider that BCT might select the ten highest-ratio projects before finding that
426 project 11 exceeds the budget remainder, at which point the algorithm looks further down the list
427 for the next affordable project (say, project 20). Optimization, in contrast, searches for the set of

428 projects that maximizes the net benefit (say, projects 1 through 9, 11, and 12). Optimization thus
429 can find that projects 11 and 12 produce greater net benefits than projects 10 and 20.

430 **Lesson 17: Intertemporal complications can limit potential cost-effectiveness.** If
431 severe enough, intertemporal issues (decision making over time) can lead to a selection of
432 parcels that is optimal today, but viewed from a broader time horizon would be suboptimal. This
433 can be referred to as myopic optimality. At a basic level, simply carrying budget remainders over
434 to future periods can improve cost-effectiveness by avoiding problems with budget remainders
435 and spending out budgets on low-priority projects. Cost-effectiveness becomes significantly
436 more complicated when the future availability of projects is uncertain or the conservation benefit
437 is time limited (extinction of a species or nonrenewability threshold). Costello & Polasky (2004)
438 assessed optimal selection in an intertemporal optimization problem and found, in part, that
439 budgets available in early periods deliver much greater benefits. Meir et al. (2004) formulated
440 the problem of dynamic budgets when benefits and project availability are uncertain and found
441 that a relatively simple, opportunistic selection strategy can outperform myopic solutions.

442 **Lesson 18: Cooperation among conservation entities can help mitigate**
443 **intertemporal issues.** This cooperation can insure against the risk that any one entity cannot
444 afford to secure an opportunistic project. One strategy common in the conservation community is
445 for a nongovernmental entity to acquire opportunistic projects and then transfer them to a
446 government agency once the governmental budget is renewed.

447

448

449

450 **4.5 Incentive Problems**

451 Conservation policy is an imperfect instrument and incentive problems may arise. Incentive
452 problems occur when, in response to a new policy, the “wrong” landowners signup (adverse
453 selection) or landowners alter their behavior in ways that work against the goals of the policy
454 (unintended consequences).

455 **Lesson 19: Adverse selection creates incentive problems that work against cost-**
456 **effective conservation policy.** Adverse selection arises because landowners typically have
457 private information about the costs of delivering conservation services. For instance, a planner
458 cannot observe how likely (or costly) it would be for a landowner to expand riparian buffers
459 without a policy incentive to do so. Voluntary conservation policy will tend to attract landowners
460 who are already most likely to deliver the conservation services, if planners do not distinguish
461 landowners by their propensity to deliver services. If owners who would already be willing to
462 supply benefits participate in a conservation program (wrong types), then some benefits are
463 erroneously attributed to the program. As programs incur costs to secure participation, they may
464 incur these costs without significant conservation gains on the ground. Likewise, the
465 conservation gains can be overstated as comparisons are not made to the outcomes that would
466 occur in the absence of the program. In these cases, the analysis that was based on observed
467 benefits and costs is invalidated. Adverse selection will be exacerbated when programs use CT
468 or reverse auctions to secure participation (Arnold et al. 2010). While the landowners’ costs are
469 not observable, the landowners most likely to offer conservation services at a low price tend to
470 be those inclined to conservation already.

471 Some recent conservation efforts have sought to address adverse selection with the
472 concept of additionality. In carbon programs, for example, landowners currently pursuing
473 sequestration (via no-tillage) are not eligible to sell carbon credits. Planners are addressing
474 complications that come with implementation, such as costly monitoring, questions of equity
475 (early adopters are sometimes punished), and complicated dynamic issues (a farmer could till
476 this year so the farmer could enter a program next year).

477 Wu & Babcock (1996) offered an early analysis of adverse selection that evaluated
478 information asymmetry (i.e., the government is unaware of landowners' costs) in the context of
479 the CRP. Their mechanism sorted landowners and achieved participation by the best attainable
480 method (this is known as second-best optimality, where the first-best outcome is unavailable
481 because of information asymmetry). An empirical study by Kirwan et al. (2005) examined
482 landowner behavior in CRP auctions and found evidence that 10–40% of the funds were
483 premiums (i.e., payments above the cost of supplying the conservation service), suggesting that
484 adverse selection may be present. Recent studies have examined ways to reduce adverse
485 selection using theory and existing program data from the United Kingdom's Environmental
486 Stewardship Scheme (Fraser 2009; Quillerou & Fraser 2010). Arnold et al. (2010) used game
487 theory and lab experiments to compare the impact of adverse selection on the cost-effectiveness
488 of various conservation policies. They found that tax instruments are more efficient than reverse
489 auctions, mechanism designs, and an absence of policy in the presence of adverse selection.

490 **Lesson 20: Unintended consequences of conservation policy may be impossible to**
491 **fully control.** In evaluating the CRP, Wu (2000) described the problem of *slippage*. Because the
492 CRP is a voluntary program and does not regulate land uses, landowners can bring previously

493 unfarmed land into production to compensate for land they enroll in the CRP. Wu found that 20
494 acres were converted for every 100 acres enrolled, thus offsetting as much as 14% of the
495 environmental benefits. Any type of incentive-based land-retirement program will likely be
496 vulnerable to this type of unintended consequence.

497 Mixed-use land markets present a related problem. For instance, some conservation
498 efforts produce benefits that accrue in part to neighboring parcels, which will increase in value.
499 If a neighboring parcel is undeveloped, its relative value for development increases, which in
500 turn raises the likelihood it will be developed or at least increase the costs of future conservation.
501 Armsworth et al. (2006) examined this phenomenon in the context of biodiversity conservation.

502

503 **5. Conclusion**

504 Although the theory of cost-effective conservation is straightforward, several decades of research
505 show that significant complications arise in real conservation planning situations. These issues
506 may partly explain planners' failure to use optimization methods. Lack of familiarity is surely
507 another. Drawing from evidence from conservation programs in the United States, this paper
508 offers a broad new synthesis of the benefits and challenges associated with cost-effective
509 conservation. The 20 lessons presented can answer many common questions about optimal
510 selection processes and can guide planners in government agencies and large conservation
511 organizations to more effectively employ their budgets.

512 The first objective of the paper was to establish a working definition of cost-effective
513 conservation as incorporating both benefits and costs that are measured commensurately with
514 money. The paper distinguished the concepts of optimization from its close relatives, such as

515 BCT, and compared the results of optimization to those of less effective selection strategies, such
516 as CT and BT. Twenty lessons were gleaned from this review regarding the problems of limiting
517 optimal selection with political constraints, using a nonmonetized benefit measures or benefit
518 indices, ignoring development risk, using incomplete cost measures, and employing cost
519 measures sequentially or as benefit indices. The paper highlighted complications associated with
520 interrelationships between benefits, issues of capitalization, and intertemporal planning. The
521 manuscript also identifies challenges that need more research guidance including incentive
522 problems and concepts of adverse selection, additionality, and slippage.

523 The implications of this synthesis are controversial, especially for those concerned about
524 monetizing environmental benefits in social terms. Because these lessons are suggested to guide
525 the selection of which conservation projects yield the most benefits and not *whether* the benefits
526 of environmental policy outweigh cost (such as the case with traditional cost benefit analysis)
527 hopefully this will not be as negatively viewed by environmental planners and policymakers.
528 Ultimately, conservation planning cannot be reduced to a simple dichotomy of cost-effective
529 versus cost-ineffective. Rather, it is a complicated process—one that is context-dependent and
530 subject to significant information problems. That said, following these lessons can help planners
531 do considerably better with their scarce resources and help lawmakers and policymakers design
532 institutions that are likely to deliver greater conservation benefits from a given budget. The
533 lessons also suggest ways for planners to determine whether the costs of acquiring improved data
534 are less than the benefit provided by improved selection. Ideally, as policy development
535 processes seek greater cost-effectiveness and then communicate prioritized needs for further
536 study, researchers can target their studies to deliver the greatest return on their efforts.

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Table 1. Summary of Twenty Lessons for Cost-Effective Conservation Planning.

Optimal Selection	Benefits		Costs	Budgets	Incentive Problems
1. Benefit targeting and cost targeting can lead to suboptimal project selection	3. Measure conservation benefits that are positive externalities.	8. Interrelationships (correlations and interactions) among conservation projects are often unobserved.	12. Include and fully account for all costs	15. Large budgets allow conservation of all projects, any selection strategy will be optimal	19. Adverse selection creates incentive problems that work against cost-effective conservation policy.
2. Efforts to distribute conservation funds evenly across political jurisdictions will tend to be suboptimal	4. Measure benefits to the public, not to experts	9. Optimal selection accounts for development risk	13. Costs should be monetized	16. Optimization improves cost-effectiveness when budget remainders are significant	20. Unintended consequences of conservation policy may be impossible to fully control.
	5. Monetize benefit measures	10. The policy process impacts the conservation benefits received	14. Sequential assessment of benefits and then costs will tend to be suboptimal	17. Intertemporal complications can limit potential cost-effectiveness	
	6. Benefits indices can lead to suboptimal project selection	11. Markets will tend to capitalize location-specific benefits		18. Cooperation among conservation entities can help mitigate intertemporal issues	
	7. Targeting conservation benefits leads to greater cost-effectiveness when thresholds are present				

Table 2: Hypothetical Example of Ranking and Benefit-Index Suboptimality

Panel A: Hypothetical Project Costs, Benefit Index, and Monetized Benefits

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>
Project ID	Costs (\$C)	Benefit Index (BI)	Monetized Benefits (\$B=7+BI)	Net Benefits (\$NB)	BI-Cost Ratio (BI/\$C)	Benefit-Cost Ratio (\$B/\$C)
A	\$1	11	\$18	\$17	11.0	18.0
B	\$2	8	\$15	\$13	4.0	7.5
C	\$3	10	\$17	\$14	3.3	5.7
D	\$5	21	\$28	\$23	4.2	5.6
E	\$1.5	1	\$8	\$6.5	0.7	5.3
F	\$1.5	1	\$8	\$6.5	0.7	5.3

Panel B: Hypothetical Project Prioritization and Selection with \$6 Budget

<i>H</i>	<i>J</i>	<i>I</i>	<i>K</i>	<i>L</i>
Prioritization	Benefit-Targeting (Ordinal/Cardinal)	Cost-Targeting (Ordinal/Cardinal)	BI-Cost Ratio (Cardinal)	Benefit-Cost Ratio (Cardinal)
1 st	D	A	A	A
2 nd	A	E	D	B
3 rd	C	F	B	C
4 th	B	B	C	D
5 th	E	C	E	E
6 th	F	D	F	F
Projects selected with \$6 budget	DA	AEFB	AD	ABC
Sum of Net Benefits (\$NB)	40	43	40	44

