

Cost-effectiveness of the Adaptive Implementation of Effective Programs Trial (Adept): Approaches to Adopting Implementation Strategies

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Abstract

Background

Theory-based methods to support clinician uptake of evidence-based practices (EBPs) are critical to improving mental health outcomes. Costs associated with effective implementation strategies can be substantial, and few have been rigorously evaluated. The purpose of this study is to conduct a cost-effectiveness analysis to identify the most cost-effective approach to deploying implementation strategies to enhance the uptake of Life Goals, a mental health EBP.

Methods

We used data from a previously conducted randomized trial to compare the cost-effectiveness of different augmentations to Replicating Effective Programs (REP) combined with external and/or internal facilitation to enhance uptake of Life Goals. REP is a low-level strategy that includes EBP packaging, training, and technical assistance. External facilitation (EF) involves external expert support, and internal facilitation (IF) augments EF with protected time for internal staff to support EBP implementation. We developed a decision tree to assess 1-year costs and outcomes for four implementation strategies: 1) REP only, 2) REP + EF 3) REP + EF add IF if necessary, 4) REP + EF/IF. The analyses used a 1-year time horizon and assumes a health payer perspective. Our outcome was quality-adjusted life years (QALYs). The economic outcome was the incremental cost-effectiveness ratio (ICER). We conducted deterministic and probabilistic sensitivity analysis (PSA).

Results

Our results indicate that REP + EF add IF is the most cost-effective option with an ICER of \$593/QALY. The REP + EF/IF and REP + EF only conditions are dominated. One-way sensitivity analyses indicate that results are sensitive to utilities for REP + EF and REP + EF add IF. The PSA results indicate that REP + EF, add IF is the optimal strategy in 30% of iterations at the threshold of \$100,000/QALY.

Conclusions

Our results suggest that the most cost-effective implementation support begins with a less intensive, less costly strategy initially and increases as needed to enhance EBP uptake. Using this approach, implementation support resources can be judiciously allocated to those clinics that would most benefit. Our results were not robust to changes in the utility measure. Further research is needed that incorporates robust and relevant utilities in implementation research to identify the most cost-effective strategies. This research advances economic evaluation of implementation by assessing costs and utilities across multiple implementation strategy combinations.

Trial registration:

ClinicalTrials.gov Identifier: NCT02151331, 05/30/2014

Contributions To The Literature

- Researchers to date have focused primarily on quantifying intervention costs; few have focused on implementation strategy costs and cost-effectiveness.
- This research focuses on advancing approaches for evaluating cost and cost-effectiveness of implementation strategies, which are provider tools/strategies to promote intervention uptake and have been understudied.
- This study is one of the first to conduct a comparative economic analysis of an adaptive implementation strategy trial, to provide useful, accessible information for communities to make well-informed decisions about resourcing implementation investments.

Introduction

Evidence-based treatments for mental health conditions, including depression, are essential to improving the public's health(1). Mental health conditions frequently co-occur with substance use disorders, and other co-occurring conditions, inciting sequelae of short- and long-term consequences(2). Mental health conditions have a significant financial toll: researchers estimated in 2008 that the annual earnings loss for serious mental illness in 2008 was \$193.2 billion(3). Collaborative care models have demonstrated effectiveness in improving outcomes among patients with mental disorders; collaborative care models such as Life Goals are designed to improve medical *and* psychiatric outcomes for persons with mood disorders through personal goal-setting aligned with wellness and symptom coping strategies and supported through collaborative care(2,4).

Most individuals suffering from depression and other mental health conditions are not receiving evidence-based practices (EBPs) such as Life Goals in community settings, resulting in poor and costly health outcomes and millions of research dollars wasted when EBPs fail to reach those most in need(5–7). Researchers increasingly recognize that EBPs must be complemented by effective implementation strategies (i.e., implementation interventions) to achieve desired public health outcomes(8). Replicating Effective Programs (REP) is an implementation strategy focused on maximizing flexibility *and* fidelity in EBP delivery(9). REP, based on the CDC's research-to-practice framework,(10) is guided by Social Learning(11) and Diffusion of Innovations Theories(12). Standard REP includes three primary components: program packaging, provider training, and facilitation. Standard REP is a low intensity, minimal cost intervention that has improved uptake of brief HIV-focused interventions(13). Researchers have also developed enhanced REP for more complex clinical behavioral interventions, which include added customization for program packaging and training, and implementation facilitation(14). Implementation facilitation (i.e. Facilitation) is a promising implementation strategy from the integrating Promoting Action on Research Implementation in Health Services (iPARIHS) framework that provides ongoing, individualized assistance for program delivery that can help enhance uptake of EBPs such as Life Goals in community clinics(14,15). Facilitation applies principles of interactive problem solving *with* practice-based knowledge to support teachers as they engage in program delivery(16,17). Individuals within (internal facilitator: IF) and outside of (external facilitator:EF) the organization can provide ongoing support for EBP implementation(14). External facilitators (EF) provide expertise, active guidance, and support for intervention delivery. Internal facilitators (IF) work in tandem with EFs to support providers in program delivery and communicate with organizational leadership and the external facilitator.

The costs associated with implementation strategies, especially multicomponent strategies such as REP+Facilitation, can be substantial. Cost is a key consideration from an organizational or system perspective when implementing new innovations(18). Understanding the resources needed to achieve desired behavioral outcomes (e.g., improved mental health) is essential to implementing and sustaining EBPs in communities(19). Most economic evaluation of implementation, however, has focused on intervention costs and not the costs of implementation strategies required to deploy and sustain them(20). Economic evaluation of implementation refers to the systematic evaluation of what outcomes a specific implementation strategy or set of competing strategies achieves and the costs of achieving them(21). Economic evaluation provides key information for decision-makers regarding implementation strategies to support and sustain EBP delivery. Organizations benefit from evidence that supports (or refutes) investment in specific strategies as an efficient use of resources, and this can help prioritize implementation efforts(18,20,22). Despite this need for practical economic information that will provide decision-makers with information on whether the cost of deploying an implementation strategy is worth the added cost (versus standard implementation or an alternative strategy), less than 10% of implementation studies include cost information and even fewer conduct comparative economic analyses(21,23). Thus, additional research is needed to advance economic evaluation of implementation as this will be instrumental in demonstrating if investment in implementation strategies is worth the additional costs(24).

Many types of cost evaluation exist, but one well suited to implementation science is cost-effectiveness analysis. Cost-effectiveness analysis assesses whether incremental benefits of one strategy versus another are sufficient to justify additional costs, and has been used to support mental health treatment-focused EBPs for clinical settings(25). Cost-effectiveness analysis (CEA) can inform decisions about resource allocation for program selection and delivery(26).

The objective of this study is to estimate the costs and conduct a cost-effectiveness analysis as part of an adaptive implementation trial comparing different implementation strategies. The goal of ADEPT (Adaptive Implementation of Effective Programs Trial) is to use a sequential multiple assignment randomized trial (SMART) design to compare the effectiveness of different augmentations to REP using EF or a combination of EF+IF on mental health outcomes among patients diagnosed with depression or bipolar disorders in community-based practices; details of the ADEPT trial are described in more detail elsewhere(14). A secondary ADEPT aim was to assess the costs for different scenarios of combining REP + facilitation (see Figures 1&2) to identify the most cost-effective implementation strategy approach. We compare 3 different implementation strategy combinations and standard REP implementation and evaluate relative cost-effectiveness to identify which implementation strategies are most cost-effective in achieving program goals. The baseline strategy was REP ONLY. Strategy 1 was REP+EF ADD IF, Strategy 2 was REP+EF/IF, and Strategy 3 was REP+EF. After 6 months of Facilitation, clinics responding to their respective implementation strategy (e.g., > discontinued. Among those who did not respond, Strategy 1 added IF, Strategy 2 continued with EF/IF and Strategy 3 continued with EF.

Methods

This study will use a simulation modeling approach using data from a previously-conducted clinical trial(14,27). Our results are reported using the Consolidated Health Economic Evaluation Reporting (CHEERS)

guidelines (28). Implementation strategies included in the model reflect implementation strategies that could be developed using data from the trial. In this study, we focus on the ADEPT community-based mental health or primary care clinics who were nonresponsive after 6 months of Replicating Effective Programs (REP) and would receive additional implementation support (i.e., Facilitation) to enhance uptake of Life Goals. Non-responsive to REP was defined as 10 or fewer patients receiving Life Goals or <50% of patients receiving a clinically significant dose of Life Goals, fewer than three Life Goals sessions (<3 out of 6), after six months (29–31). Eligible sites had at least 100 unique patients diagnosed with depression and could designate at least 1 mental health provider to administer individual or group collaborative care sessions for patients. The study was approved by local institutional review boards (IRBs) and registered under clinicaltrials.gov (identifier: [NCT02151331](https://clinicaltrials.gov/ct2/show/study/NCT02151331)).

Modeling Approach

Using data from the ADEPT trial, we designed a cost-effectiveness study to evaluate three strategies that could be implemented to support the uptake and clinical effectiveness of Life Goals. These strategies do not exactly match the arms in the clinical trial because our goal was to evaluate the optimal strategies for Phase II (or non-responders). We developed a decision tree to assess 1-year costs and outcomes for different intervention strategies following 6 months of REP among non-responsive sites. Interventions included in the model (see Figure 2) were: 1) REP+EF for 12 months, 2) REP+EF for 6 months, add IF for 6 months, 3) REP+EF for 6 months, no implementation strategy for 6 months (responder), 4) REP+EF/IF for 12 months, 5) REP+EF/IF for 6 months, no implementation strategy for 6 months (responder). Probability of non-response to the implementation strategies in the model was based on observed non-response rates in the study, which remained consistent across each phase at approximately .09. Sites who responded to their assigned implementation strategy after 6 months discontinued the strategy. The analysis uses a 1-year time horizon and assumes a health sector perspective. Parameter inputs were derived using primary data from ADEPT.

Costs

Implementation strategy costs include personnel costs for intervention training and delivery, time spent training, supervising, providing technical assistance and Facilitation, training compensation (e.g., pay during non-work hours), time costs of assisting with intervention delivery for clinical providers. Facilitation costs are based on the facilitation logs. The study EF and site IFs logged their tasks, categorizing mode, personnel interaction, duration, and the primary focus of each task. We calculated costs based on time spent by hourly wage plus fringe rates for facilitators. As there was one EF employed by the study team, we used the EF hourly wage + fringe. For the IFs, training, and background (and thus costs) varied. We based the IF salary and fringe rates on current rates for LMSW (Licensed Masters of Social Work) professional using Bureau of Labor Statistics data, as many of the IFs were LMSWs. Non-labor costs included costs of the curriculum (manual and materials), and travel costs (20,32). As we anticipated differences in uptake, that is the number of patients receiving Life Goals by condition, we calculated the total site-level cost per strategy (the level of randomization) and divided by the number of patients in that implementation strategy condition. The number of patients per condition was obtained from site-level records. Costs were collected in 2014 and adjusted to US 2018 dollars using the Consumer Price Index (33). A summary of cost parameters is provided in Table 1. We report summary statistics for implementation costs with 95% confidence intervals. We estimated the costs of REP using the available cost data to obtain a comprehensive assessment of total implementation intervention costs, plus the costs of facilitation activities in each condition (EF and EF/IF).

Health Outcomes

Quality-adjusted life-years (QALYs). To develop a preference-based health utility measure for the current study, we mapped the SF-12 (which was collected as part of the patient-level evaluation in the ADEPT trial) to the EQ-5D, a multi-attribute utility instrument, using an established algorithm developed by Franks and colleagues(34). The EQ-5D yields interval-level scores ranging from 0 (dead) to 1 (perfect health). This mapping provides a health utility measure for each health state experienced by patients in the study and can be used to calculate quality-adjusted life years, the preferred measure for health benefits used in cost-effectiveness analysis.

Data Analytic Approach

We used a decision-tree model to compare the cost-effectiveness across different scenarios for combining REP + facilitation for the Life Goals EBP (See Figure 2). The time horizon for this analysis was 12 months as this is the duration of the trial phase of the study. In this analysis, we adopted a health system/payer perspective. This narrower perspective stands in contrast to the full, societal perspective, which incorporates all relevant costs and benefits and is recommended for most economic evaluations(35). While this is a narrower perspective can potentially ignore important costs or benefits from the broad societal standpoint, it has the practical value of explicitly addressing the budgetary concerns of payers. Thus, this approach fits well with implementation science contexts where financial factors are often central to whether programs and services are adopted and sustained(36).

Assumptions were made on the psychometric properties of the outcome measures, the effectiveness of the Life Goals intervention, and the reliability of time reporting by the Facilitators. We test these assumptions in the sensitivity analyses by varying the costs and outcomes related to each intervention condition at low and high values (95% confidence interval). To address missing data on our utility (outcome) measures, we employed an inverse probability weighting (IPW) approach(37).

We estimated per-patient costs and QALYs following the 12-month Trial Phase for each implementation strategy sequence. We calculated the per-patient cost by dividing the total costs per condition by the number of patients in each condition. To compare interventions, we divided net incremental costs (net increase in costs from REP+EF/IF versus REP+EF, for example) by incremental effectiveness (net increase in QALYs in REP+EF/IF versus REP+EF groups, for example) to calculate the incremental cost-effectiveness ratio for patient-level outcomes across the conditions. We conducted a one-way sensitivity analysis on all input parameters listed in Table 1 to create a tornado diagram using Net Monetary Benefits (NMB). We used NMB as this facilitates multiple comparisons, as in the current study, and incremental cost-effectiveness ratios (ICERs) are less suitable with more than 2 comparators(38). This sensitivity analysis evaluates how costs and incremental cost-effectiveness is affected by variations in key parameters(26). When available, we based upper/lower bound estimates on the 95% confidence intervals. We also conducted a probabilistic sensitivity analysis (PSA). PSA characterizes uncertainty in all parameters simultaneously, reflecting the likelihood that each model parameter takes on a specific value and provides information on overall decision uncertainty based on parameter uncertainty(39). We conducted 1,000 model simulations to quantify the probability that the implementation strategy is cost-effective for a range of thresholds of willingness-to-pay(40).

Results

Results of Base Case Analysis

Base case results are presented in Table 2 and a plot of cost-effectiveness across implementation strategies is depicted in Figure 2. Our base case analysis results indicate REP ONLY is the least expensive. REP+EF, ADDIF has an ICER of \$593/QALY. REP+EF had higher QALYs than REP alone, but the QALYs were not as high as REP+EF, ADD IF, and it was higher cost than REP+EF, ADD IF. REP+EF/IF had higher costs and lower QALYs than REP ONLY.

Sensitivity analysis

To test our model assumptions, we conducted sensitivity analyses on all parameters. In the tornado diagram (see Figure 3), we found the results were sensitive to the following variables (in order): utility of individuals in the REP+EF,ADD IF arm at 12 months (Phase III), the utility of individuals in the REP+EF arm at 6 months (Phase II), the utility of individuals in the REP+EF arm at 12months for responders (Phase III), and utility of individuals in the REP+EF only arm at 12 months (Phase III). We then conducted threshold analyses for each of the most sensitive parameters. We found that at utility values below .44 for REP+EF,ADD IF at 12 months, the value of REP+EF, ADDIF is no longer cost-effective and REP+EF becomes the most cost-effective choice. We also found that at utility values above .57 for REP+EFat 12 months, REP+EF ADD IF is no longer the most cost-effective option and REP+EF becomes the most cost-effective choice.

In addition to the deterministic sensitivity analyses, we also conducted probabilistic sensitivity analysis. The results indicate that the intervention with the best results in terms of utility would be most preferred. The willingness-to-pay threshold was not important (unless using a \$0 willingness-to-pay threshold). REP+EF, ADD IF is the optimal strategy in about 30% of iterations, REPOONLY is the optimal strategy in 31% of the iterations, and REP+EF/IF the optimal strategy in 22% of the iterations.

Discussion

Effective implementation of EBPs for mental health treatment in communities is critical to improving the public's health. Most individuals suffering from depression and other mental health conditions are not receiving evidence-basedpractices (EBPs) such as Life Goals (LG) in community settings, resulting in poor and costly health outcomes and millions of research dollars wasted when EBPs fail to reach those most in need(5–7). Implementation strategies arekey to improving uptake of EBPs in communities and achieving public health objectives of evidence-based treatments such as Life Goals. Implementation strategies, however, vary in intensity and cost. more research is needed on applying these strategies with consideration of the economic impact, given that community clinics often have limited—and carefully allocated—resources to promote EBP uptake(41).This research is vital to bridging the research-to-practice gap, but economic evaluation of implementation strategies remains understudied(41).This study is one of the first to investigate the cost-effectiveness of implementation strategies as part of an adaptive trial. Adaptive trials are an effective way to accelerate research to practice translation by simultaneously evaluating multiple strategies and combinations of strategies, based on clinics' needs.

We found that, overall, REP + Facilitation in its various permutations is a relatively low-cost implementation strategy. Identifying the costs and potential utilities for each REP + Facilitation combination can help decision-

makers with resource allocation for implementation. Understanding the resources needed to achieve desired behavioral outcomes (e.g., reduced ATOD use) is essential to implementing and sustaining EBIs(19). Also, we found that REP+EF, ADD IF may be the most cost-effective implementation strategy. But these results are still uncertain based on highly variable quality-of-life assessments by participants. Although researchers have debated if a step-up versus step-down approach to evidence-based clinical treatment is most effective, the optimal approach for implementation strategies to enhance the uptake of these treatments is also unclear. Our results are consistent with other clinical research that suggests a step-up strategy is a more cost-effective approach(42). This information will support organizations in making informed decisions by providing evidence that supports (or refutes) investment in specific implementation strategies as an efficient use of resources and thus can help prioritize implementation efforts.

We also found that stepping up non-responsive sites immediately to REP+EF/IF the most intensive and costly strategy (at the site level), was not as likely to be cost-effective. This may be for several reasons. First, EF alone may be sufficient for community clinics to effectively implement the Life Goals intervention and, thus, in most cases IF may not be necessary(27). Second, many sites had difficulty identifying an internal facilitator. Subsequent analyses into time data indicate that the time to identify an IF was 69 days. This suggests that many sites that had an IF did not have one for the first 2 months of the evaluation period. These results also indicate that community clinics may have a limited capacity to identify and effectively utilize an IF. Finally, we may have had more favorable results with the REP+EF, ADD IF condition during Phase II as the EF was able to work with the clinic on their barriers to uptake immediately and may have been working with several, versus a single staff member. The results of our sensitivity analyses indicate that the cost-effectiveness may vary depending on specific parameters. In particular, at the lower end of the 95% confidence interval for the REP+EF, ADD IF utilities at 12 months and at the lower end of the 95% CI for REP+EF utilities at 12 months the relative cost-effectiveness changes. At the lower bound, for example, REP+EF, ADD IF may have worse cost-effectiveness than other REP+Facilitation and/or REP ONLY scenarios. Given that the utility of REP ONLY at 12 months was also an influential parameter, our results also indicate

that clinic implementation support and this patient outcomes may be especially critical following initial uptake.

Our results were highly dependent on the assessment of utility. The utilities were highly variable and uncertain across the different intervention arms. This had a strong influence on the overall assessment of cost-effectiveness. Further research is needed that incorporates robust and relevant utilities in implementation research to identify the most cost-effective strategies.

Limitations

We adopted a health payer perspective, which may not account for other relevant costs if considering the societal perspective. This may include indirect costs such as patient time, costs of hospitalization, or other treatments or lost productivity. Yet, this narrower perspective has the practical value of explicitly addressing the budgetary concerns of payers and fits well with implementation science contexts where financial factors are often central to whether programs and services are adopted and sustained(36). We did not have additional information on estimates of REP costs to vary parameters and these cost estimates primarily relied on research team recall. There may be additional costs not included in the estimates that may have implications on the CEA results. In addition, additional information to vary specific parameters may also help inform those parameters

that are most influential on our estimates. In our CEA analyses, however, all groups had REP costs incorporated into total costs, so this is unlikely to influence the CEA results across the REP+Facilitation permutations. We did not have a direct measure of QALYs and thus our utility estimates may be especially susceptible to measurement error. A notable amount of research has been done, however, on mapping the SF-12 components to the EQ-5D thus reducing the likelihood of error as a result of the mapping process. Next, we had a notable amount of missing patient-level survey data, increasing the likelihood of biased estimates. We did, however, attempt to reduce this bias using inverse probability weighting.

The current study would benefit from a mixed methods approach, specifically a sequential design, to obtain qualitative data following the quantitative data collection to better understand challenges to utilizing IF/EF. Finally, our study has a limited time horizon. Although we estimated incremental QALYs gained based on the survey results, a longer time horizon would provide additional information for a longer-term return-on-investment.

Conclusions

Our study has several practice implications. First, our results support using a step-up strategy for implementation support for sites that are slow to implement as a cost-effective approach to enhancing uptake and clinical outcomes. Second, our results provide information for decision-makers and community health clinic leadership on the costs and relative benefits of using various implementation strategies to improve clinical outcomes. Third, our results support the need for further cost-effectiveness research and incorporating robust utility assessments in community health clinics to provide evidence that will support or refute investments in specific strategies. Finally, our results point to the need for mid-program utility evaluation for both effectiveness and cost-effectiveness to make accurate determinations of the most efficient implementation strategy approach.

List Of Abbreviations

EBPs: Evidence-based practices

REP: Replicating Effective Programs

CDC: Centers for Disease Control and Prevention

iPARIHS: integrating Promoting Action on Research Implementation in Health Services

IF: Internal facilitation/facilitator

EF: External facilitation/facilitator

VA: Veteran's Administration

CEA: A cost-effectiveness analysis

ADEPT: Adaptive Implementation of Effective Programs Trial

SMART: sequential multiple assignment randomized trial

LMSW: Licensed Masters of Social Work

QALYs: Quality-adjusted life-years

IPW: inverse probability weighting

NMB: net monetary benefits

ICER: incremental cost-effectiveness ratio

PSA: probabilistic sensitivity analysis

Declarations

Ethics approval and consent to participate

The study recruited primary care and community mental health sites across the state of Michigan. Prior to the first randomization, sites that agreed to participate were asked to provide the names of at least 20 patients suitable for LG. These patients were then contacted by the study survey coordinator to confirm eligibility and obtain patient consent. The study was approved by local institutional review boards and registered under clinicaltrials.gov.

Trial registration: ClinicalTrials.gov Identifier: NCT02151331, registered 05/30/2014, <https://clinicaltrials.gov/ct2/show/NCT02151331>

Consent for publication

Not applicable.

Availability of data and materials

Deidentified data are available on request.

Competing interests

None.

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Authors' contributions

AE conceived of the study, conducted the analyses, and drafted the manuscript. DH and LP aided with design and interpretations of cost-effectiveness models & results provided critical input on the full manuscript. SS Oversaw data collection and cleaning, aided with interpretations of models & results, drafted sections of the

manuscript, provided critical input on the full manuscript. AK led the data collection, project conception, and provided input on study conception. All authors worked on the interpretation of data, and critical review and approval of the final manuscript.

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Tables

Table 1: Model inputs

Parameter	Base	Low	High	Source
Costs				
Cost of REP (Phase I)	588.95	0	558.95	Time & resource tracking, study staff
Additional cost of EF (Phase II)	32.70	32.39	33.01	Time logs
Additional cost of EF (Phase III)	17.55	1.22	30.84	Time logs
Additional cost of EF and IF (Phase II)	31.23	28.51	30.49	Time logs
Additional cost of EF and IF (Phase III)	6.35	3.15	9.27	Time logs
Probabilities				
Probability of response after Phase II(6 months) with REP+EF	0.095	0	0.095	Site response data
Probability of response after Phase II (6 months) with REP+EF/IF	0.091	0	0.091	Site response data
Utilities^{a, b}				
REP only	.475	0.43	0.521	Patient survey
EF (Phase II: 0-6 months)	.497	0.42	0.573	Patient survey
EF non-responding site (Phase III: 6-12 months)	.446	0.306	0.586	Patient survey
EF responding site (Phase III: 6-12 months)	.721	0.533	0.909	Patient survey
EF add IF (Phase III: 6-12 months)	.568	0.392	0.566	Patient survey
EF and IF (Phase II: 0-6 months)	.463	0.362	0.564	Patient survey
EF and IF non-responding site (Phase III: 6-12 months)	.479	0.392	0.566	Patient survey
EF and IF responding site (Phase III: 6-12 months)	.372	0.184	0.559	Patient survey

^ainverse probability weighting to account for missing data with weights estimated from a logistic regression model for predicting non-response;^bEQ-5D calculated using mapping algorithm from components of the SF-12

Table 2: Base case analysis results

Condition	Cost per patient	Incremental cost	Effectiveness	Incremental effectiveness	ICER ^a	NMB ^b	
Utility: QALYS							
Baseline (REP only)	558.95	0	0.47	0	0	46911.05	
REP + EF/IF	625.95	37.00	0.47	0.01	-4174.67	45987.68	dominated
REP + EF, add IF if needed	627.40	38.45	0.54	0.06	593.42	53351.18	
REP + EF only	637.53	10.13	0.48	-0.06	-183.61	47822.00	dominated

^aICER: incremental cost-effectiveness ratio, ^bNMB: net monetary benefit. Willingness-to-pay threshold \$100,000/QALY.(43)

Figures

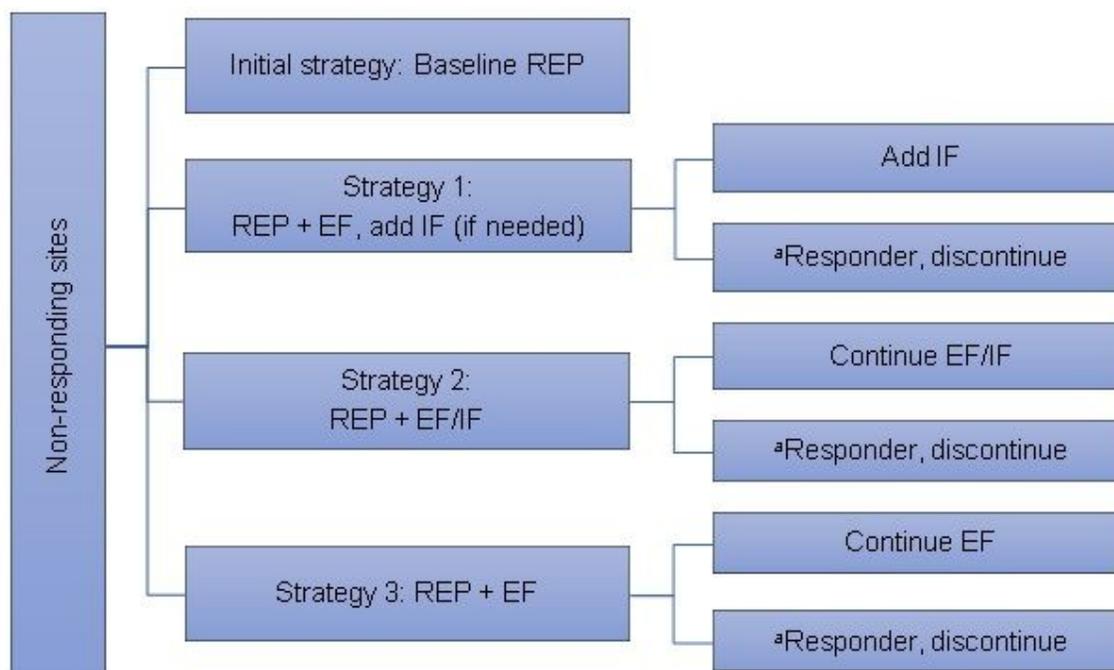


Figure 1

Decision tree of the ADEPT trial. aSites that responded to the implementation strategy after initial 6 months of the Trial Phase: either <10 patients receiving Life Goals or >50% of patients receiving Life Goals had ≤ 3

sessions, min dose for clinically significant results. ∅ indicates that the site responded to the implementation strategy and discontinued the strategy during the second 6 months.

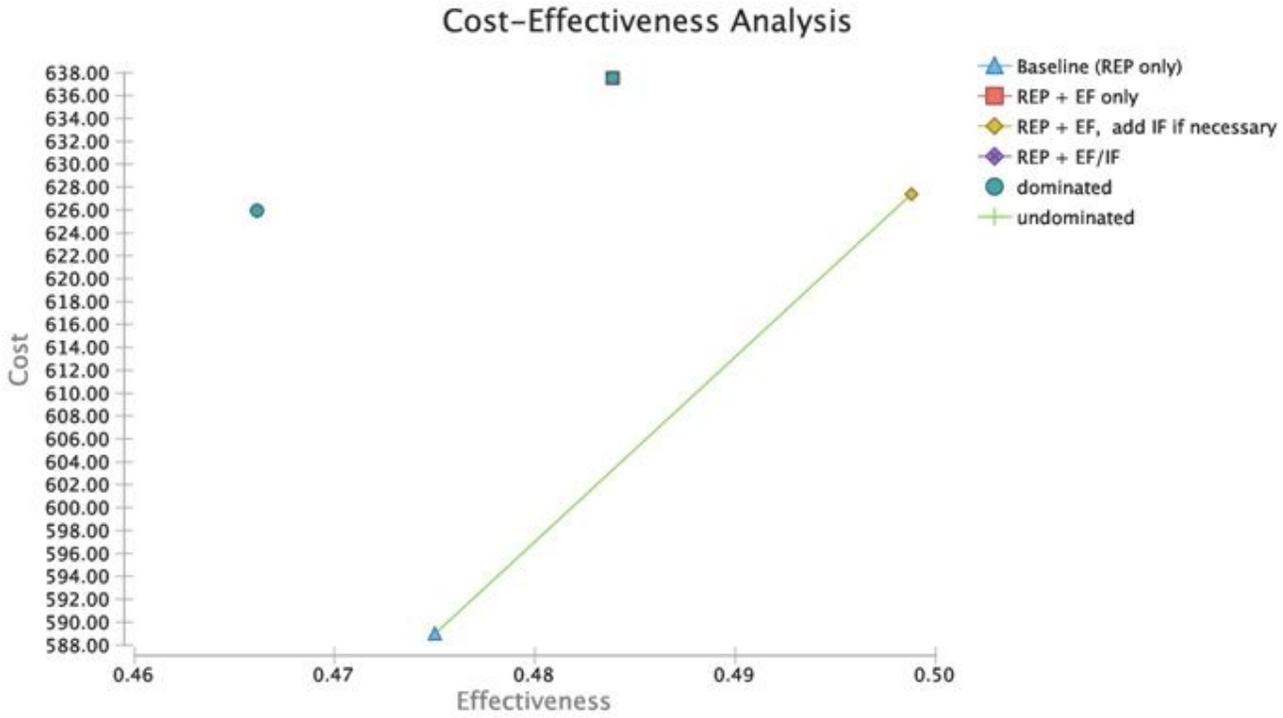


Figure 2

Cost-effectiveness plane, organization/payer perspective

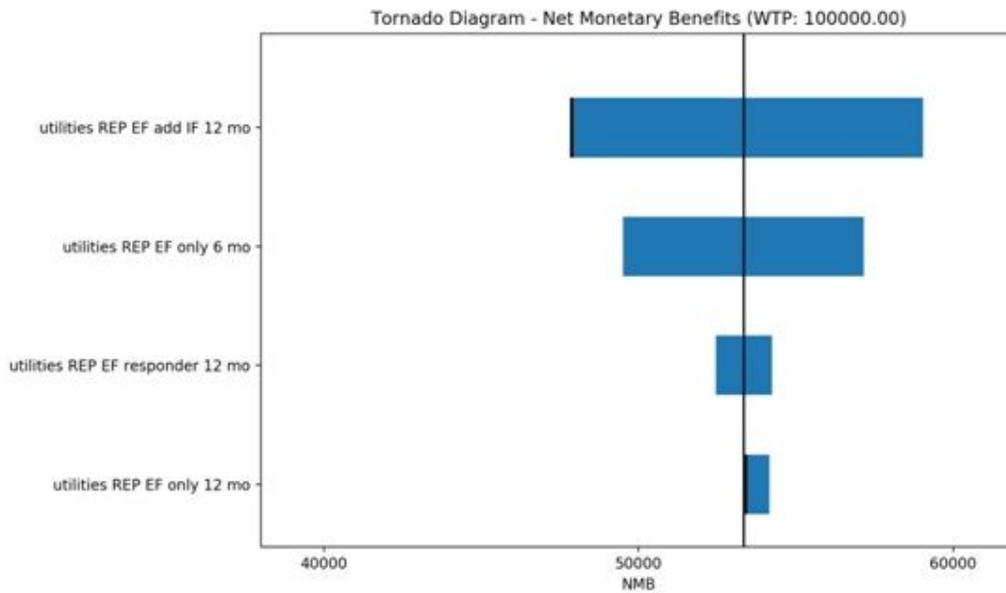


Figure 3

Tornado diagram showing one-way sensitivity analyses for the base case with the most sensitive parameters. All parameters were evaluated. Thick vertical black lines on the ends of the utility bars indicate values at which

the initial preferred option is no longer cost-effective.

Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- [checklistcompleted.pdf](#)