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Perspectives on Quality

How do we learn about improving health care:
a call for a new epistemological paradigm

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Abstract

Purpose: The field of improving health care has been achieving more significant results in outcomes at scale in recent years. This has raised legitimate questions regarding the rigor, attribution, generalizability and replicability of the results. This paper describes the issue and outlines questions to be addressed in order to develop an epistemological paradigm that responds to these questions.

Questions: We need to consider the following questions: (i) Did the improvements work? (ii) Why did they work? (iii) How do we know that the results can be attributed to the changes made? (iv) How can we replicate them? (Note, the goal is not to copy what was done, but to affect factors that can yield similar results in a different context.)

Next steps: Answers to these questions will help improvers find ways to increase the rigor of their improvements, attribute the results to the changes made and better understand what is context specific and what is generalizable about the improvement.

Key words: improvement, learning, complex adaptive systems, implementation, delivery

Introduction

This article raises an important issue in the field of improving health care today: ‘How do we learn about improving healthcare, so that we can make our improvement efforts more rigorous, attributable, generalizable and replicable?’ (See Fig. 1). Intended to outline what we know and what remains to be answered, the article represents a synthesis and analysis of the knowledge and experiences of the authors and reviewers. Although we acknowledge that the improvement community’s opinions differ regarding the definition of certain terms, this article is framed by the working definitions found in Table 1. The paper concludes with key questions that need to be addressed in order to advance the field of healthcare improvement.

Background

The World Health Organization (WHO) describes quality care as care that is effective, efficient, accessible, acceptable, patient-centered, equitable and safe [1]. Yet, much of the care received in high-, middle- and low-income settings does not meet the WHO criteria, often due to the complexity of healthcare—which is why we need to improve healthcare [2]. For the purposes of this paper, we will use the following working definition for improving healthcare: ‘The actions taken to ensure that interventions established to be efficacious are implemented effectively every time they are needed.’

The multi-level structure found in healthcare settings exists as interconnected sets of autonomous healthcare providers, teams and units within healthcare organizations, nested within health systems.
Multiple approaches are used for improving healthcare. [13] They generally focus on analyzing and iteratively testing changes to the processes of care delivery, with the objective of improving results of the process and outcome levels. The activity essentially consists of re-organizing care delivery to ensure that evidence-based care is implemented reliably. The changes to be tested occur in a continuous cycle of testing adaptation and learning and can include, but are not limited to:

- Work with multidisciplinary teams of healthcare providers, engaging them in testing care process re-design to ensure the reliable implementation of what has been shown, through evidence, to work
- Review health workers’ competencies and their workload and test changes to improve what is done by whom at what point in the care delivery process
- Work with patients, their families, and communities to understand and test changes, ascertain that services meet their needs and expectations, and engage them in their care
- Test changes to correct for any supply-chain issues or inputs that may be necessary
- Work with and engage leadership structures to test system-level changes, as needed
- Work on changing norms and policies to institute re-designed care processes
- Work and test changes on any other matters, as the needs arise
- Throughout the activity, data are plotted in real time, using time-series charts to track the impact of the changes. Changes tested are usually annotated on the time-series chart to aid in the interpretation of results. Based on this analysis, new changes are tested and tracked until the aim is achieved.
- Facilitate experiential learning through the use of iterative cycles of implementation and evaluation, and harvest and share this experience through peer-to-peer learning sessions.

Figure 1 Codifying improvement.

The dynamic human interactions between and among healthcare workers, patients, managers, payers and other actors and the variety of social, cultural, economic, historical, political and other factors within this multi-level structure put improving health care squarely in the arena of complex adaptive systems [3]. Complex adaptive systems are comprised of individuals who learn, self-organize and evolve in response to changes in their internal and external environment, and interrelate in a non-linear fashion to accomplish their work and tasks [3, 4]. These complex adaptive systems, in which implementers operate, present unique challenges. They cannot be reduced to their component parts without the risk of negating the original intention of the intervention, therefore interventions must be implemented in a contextually adaptive manner in order to effectively work in any given system.

The issue

The field of improving health care has evolved to address the complex, interdependent, systemic nature of healthcare challenges, particularly through the use of adaptive, iterative testing and implementation of changes, and by empowering teams to use data in real time to do so. Teams and individuals who learn in these complex, adaptive systems use real-time data to assess whether introduced changes lead to an improvement in the outcome of interest—after which, they institute, adapt or discard these changes as needed, in a continuous cycle of testing and learning. This process is demonstrated in the example in Fig. 2.

Improvements like these have generally relied on the use of time-series charts, which plot indicators of the improvement sought against frequent (daily or weekly) time intervals before, during and for a time period following the introduction of changes. At times, they have also used control limits, but as a rule, improvement has relied on analytic statistics, as opposed to enumerative statistics, to establish significance [5].

Improving healthcare engages all key stakeholders involved in the outcome of interest, including healthcare providers directly involved in delivering care, together with their supervisory and support structures, as required. This process can also include patients, families, communities and other stakeholders. By meaningfully engaging all key stakeholders, health teams are able to address systems and micro- and macro-political human-factors issues [6] and build ownership through restructuring processes. In addition, the adaptive, iterative nature of testing and learning allows the teams to implement changes appropriate to the local context and responsive to emergent and shifting dynamics among different actors and at various levels of the system.

Over the past two decades, the application of modern improvement methods has expanded beyond the administrative processes in facilities, where it was first applied (for example, to reduce waiting times) to include clinical improvements (like reducing the incidence of hospital-acquired infections) and more significant improvements in health care (like reducing secondary complications and decreases in mortality); and, moreover, achieving these outcomes and results at scale [7, 8]. This evolution has generated greater interest in improvement and simultaneously raised more questions regarding the validity, rigor, attribution, generalizability and replicability of the results.

In response to these legitimate questions, there have emerged two schools of thought. One school of thought calls for continuing to accept analytic statistics as sufficient evidence for improvement [5]. The other calls for subjecting improvement to the same enumerative statistical methods used in clinical research, such as the use of randomized control trials.

Discussion

The use of randomized controlled trials, the gold standard for clinical research, has been limited in the field of improving health care. In part,
Improving healthcare The actions taken to ensure that interventions established to be efficacious are implemented effectively every time they are needed.

Complex adaptive system (CAS) A CAS is comprised of individuals who learn, self-organize, evolve in response to changes in their internal and external environment, and inter-relate in a non-linear fashion to accomplish their work and tasks [3, 5].

Time-series charts Time-series charts are a graphical presentation of an indicator over time and are a common tool used to track continuous-quality-improvement data. Statistical process control was developed by Shewhart [28] and popularized by Deming [29] as a way to apply statistical methods to distinguish between natural or ‘common cause variation’ in a process (i.e. random variation that could be computed from a statistical model based on the Gaussian, Poisson or binomial distribution), versus ‘special cause variation,’ which is variation that follows certain defined patterns (e.g. 2 out of 3 successive points >2 SD from the mean) [30]. This adaptation of traditional statistical hypothesis testing into easily visualized tests is a practical way to add rigor to the interpretation of time-series charts.

Figure 1. Percent of patients whose nutritional status is assessed using mid-upper arm circumference

This knowledge may be generalizable or may need further adaption for systems that allow for rapid learning in order to build this knowledge. This knowledge may be generalizable or may need further adaption for a different context [14].

A 2015 literature review by Portela et al. [15] provided a useful overview of various methods that can be used to learn about improving health care and describes the strengths and weakness of various approaches that range from more traditional, experimental designs to quasi-experimental designs, as well as systematic reviews, program and process evaluations, qualitative methods and economic evaluations. The authors note that the dichotomy between designs classified as practical (‘aimed at producing change’) and those classified as scientific (‘aimed at producing new knowledge’) may be a false one, and that the field should find ways to optimize rigor and generalizability of studies, without compromising the importance of adaptability and context.

On this note, Davidoff et al. and other authors call for the demystification of the use of theory in improving health care [16], including more use of theory a priori to better understand how and why an improvement occurs and provide insight into the so-called ‘black box’ of improvement. Parry et al. provide a guide for a formative, theory-driven approach to evaluating improvement initiatives by
defining three improvement phases for initiatives (innovation, testing, and scale up and spread), each defined by the degree of belief in the intervention, and each requiring a different evaluation approach [17].

Campbell et al. [18] also emphasize the importance of phasing; for example, in the use of randomized controlled trials for complex interventions. The authors describe a more flexible approach to the 2000 Medical Research Council framework [19] on this topic by considering an iterative, stepwise approach to building understanding of the context of the problem, intervention and evaluation, in order to obtain meaningful information from randomized, controlled trials of complex interventions.

Braithwaite et al. urge improvement teams to learn from the way clinicians adjust their behavior with their concept of Safety II: ‘we must understand how frontline staff facilitate and manage their work flexibility and safely, instead of insisting on blind compliance or the standardization of their work’ [20].

Others have called attention to different ways to integrate adaptability with the fixed concept of randomized, controlled, trial interventions (delineating between standardization of, but not the form of, the intervention) [21] and the need for earlier pilot testing with iterative learning and non-linear evaluation processes, in order to more fully understand complex adaptive systems [22].

Furthermore, as a social intervention, improving health care is complex and, therefore, difficult to understand, design, implement, reproduce, describe and report. Frameworks, such as the Consolidated Framework for Implementation Research, developed by Damschroder et al. [23], provide guidance for ways to evaluate complex adaptive systems by classifying them into various domains, which describe various intervention and contextual factors relevant to the evaluation. The Template for Intervention Description and Replication (TIDieR) guidelines developed by Hoffmann et al. [24] and the Standards for Quality-Improvement Reporting Excellence (SQUIRE) Guidelines [25] provide a guide for improved reporting of interventions in order to improve the completeness of reporting and aid replicability [26].

Advancing the field: issues that need to be addressed

The work presented by the above-mentioned authors is rooted in the ideas of showing that improvement has occurred, why it has occurred, and how to better learn about improvement to make it happen more effectively and in other contexts. There is evidence to support these objectives and we must build upon this to move improvement forward. These are also the themes that lead us to the next level of questions for improvement.

In order to learn about improving health care, there is a need for thought leaders in improving health care, as well as from related fields that also use complex adaptive systems; researchers; and others to come together to co-develop a robust framework that has widespread support and that reflects the diverse, nuanced ways we learn about improving health care. Many questions remain to be answered. These include, but are not limited to:

- If other factors are also affecting the results, how do we know what part is attributable to the changes we are making?
- Why did the changes which yielded improvements work, and how?
- If we obtain good results, and have documented the changes which yielded them, to what extent can we implement them elsewhere with the same fidelity and expect to get similar results?
- What elements of an improvement are transferable, and what adaptations are needed?
- How can we incorporate the effects of local context into the improvements?
- How should we design improvement efforts to answer different learning objectives?
- How do we optimize data collection that simultaneously serves to drive quality improvement, inform evaluation efforts, and fulfill performance reporting requirements?

Answers to these questions will necessarily lead to multiple study types, depending on what we want to learn, in what context, and for what purpose.

A phenomenon that is relevant to learning in general which also equally applies to improving health care, is that successful work which produces results tend to get published. Learning comes from not only what has worked and why, but also what has not worked and why not. In order to enhance learning we need to be deliberate about studying not only success, but also failures. The changes which failed, why they did not work and under what circumstance [10, 27].

Why/so what

A new framework for how we learn about improvement will help in the design, implementation and evaluation of improving health care to strengthen attribution and better understand variations in effectiveness through reproducible findings in different contexts. This will in turn allow us to understand which activities, under which conditions, are most effective at achieving sustained results in health outcomes.

Conclusion

The complexity of health care requires a more rigorous approach to advance our understanding of methods for learning about improving health care. Additionally, the greater use of robust qualitative, quantitative and mixed methods is needed to assess effectiveness—not merely to demonstrate if an intervention works, but why and how it works—and to explore the factors underlying success or failure.

Key questions to examine further include how to strengthen the rigor of the improvement; increase attribution of results to the changes tested; provide better balance to the often opposing needs of improving fidelity of the intervention, versus allowing for adaptation; make conclusions that are generalizable, but that also respond to the local context; and account for political considerations in improvement activities. This can lead to an improved epistemological paradigm for improvement.

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