An Adult Immunization Best Practices Learning Collaborative: Impact, Scale Up, and Spread

Elizabeth L. Ciemins, PhD, MPH, MA,1 Michelle Jerry, MS,2 Jill Powelson, DrPH, MPH, MBA, RN, CPC,1 Erin Leaver-Schmidt, MPH,1 Vaishali Joshi, BS,1 Earlean Chambers, RN, MS,1 Danielle Casanova, MBA,1 John W. Kennedy, MD,1 and Jerry Penso, MD1

Abstract

The research objective was to rapidly scale up and spread a proven learning collaborative approach (intervention) for adult vaccination rates for influenza and pneumococcal disease from 7 to 39 US health care organizations and to examine improvement in adult immunization rates after scale-up. Comparative analyses were conducted between intervention and nonintervention propensity score-matched providers on vaccination rates using a difference-in-differences approach. Qualitative data, collected during site visits and in-person and virtual meetings, were used to enhance understanding of quantitative results. In 2017–2018, an analysis of a subset of sites (n=9) from 2 intervention cohorts (~20 sites each) demonstrated greater improvement than their matched providers in pneumococcal vaccinations (PV) for patients ages ≥65 years (treatment effect range: 1.4%-3.7%, P<0.01) and PV for high-risk patients (eg, with immunocompromising conditions) aged 19–64 years (0.8%-1.6%, P<0.01). Significant effects were observed in one of the study cohorts for PV for at-risk patients (eg, with diabetes) aged 19–64 years (1.7%, P<0.01), and influenza vaccination rates (2.4%, P<0.001). Individual health systems demonstrated even greater improvements across all 4 vaccinations: 9.5% influenza; 8.7% PV ages ≥65 years; 11.8% PV high-risk; 16.3% PV at-risk (all P<0.01). Results demonstrated that a 7-site pilot could be successfully scaled to 39 additional sites, with similar improvements in vaccination rates. Between 2014 and 2018, vaccination improvements among all 46 groups (7 pilot, 39 in subsequent cohorts) resulted in an estimated 5.5 million adult vaccinations administered or documented in 27 states.

Keywords: learning collaborative, prevention, immunizations, vaccinations, disseminations, scale up

Introduction

According to the Institute for Healthcare Improvement (IHI), the key to a best practice becoming a common practice is the ability of providers and health systems to rapidly spread new ideas and successful innovations.1 IHI’s Framework for Spread1 is a useful model for identifying key components for the spread of best practices and innovations and aligns with AMGA’s efforts to spread its Adult Immunization Best Practices Learning Collaborative to 39 health care organizations. This was accomplished by bringing health systems and their providers and staff together, both in-person and virtually, in an iterative learning collaborative format. This article demonstrates the scale-up and spread that followed a pilot learning collaborative in efforts to increase adult immunization rates nationwide.

According to the Centers for Disease Control and Prevention (CDC), introduction of pneumococcal conjugate vaccines (PCV) in the United States reduced invasive pneumococcal disease by 56% among adults aged 19–64 years and by 61% among adults aged ≥65 years between 1998–2015.2 For the last flu season on record (2017–2018), adult influenza vaccines prevented an estimated 7 million flu illnesses, 109,000 flu hospitalizations, and 8000 flu deaths.3 Pneumococcal vaccination (PV) also reduces the likelihood of prolonged hospitalization among adults, especially those at high risk of

1AMGA Analytics, Alexandria, Virginia, USA.
2Optum Analytics, Cambridge, Massachusetts, USA.

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disease. A study among patients at high risk for cardiovascular disease found that individuals who received vaccinations for influenza had significantly lower risks of having adverse cardiac events, a leading cause of death in the United States.

Despite the strong evidence base for adult immunization, millions of illnesses and tens of thousands of deaths occur every year from vaccine-preventable diseases. In 2017, CDC estimated 450,000 hospitalizations related to pneumococcal pneumonia and 959,000 hospitalizations resulting in 79,400 influenza-related deaths (2017–2018 flu season). As many as 70% of influenza hospitalizations and 70%-85% of all influenza-related deaths were in older adults age ≥65 years. Estimates of the health care costs associated with influenza in adults are $5.8 billion per year, while pneumococcal disease is estimated to cost another $1.9 billion annually.

Notwithstanding the pressure to provide more efficient, affordable, value-based care, gaps in immunization among adult populations persist. In the 2017–2018 flu season, CDC estimated the percentage of adults aged 18–64 vaccinated for influenza in the United States was 31% (Healthy People 2020 goal = 70%) and the percentage of adults aged ≥65 years vaccinated for influenza was ~60% (Healthy People 2020 goal = 90%). Although there has been significant progress in reducing the gap in influenza vaccination for high-risk adults with compromised health conditions, still only 39% aged 18–64 received an influenza vaccine. The greatest gap in PV was among at-risk adults ≥19 years, with 24.5% ever having received a pneumococcal vaccine. The economic and disease burden associated with vaccine-preventable illnesses suggests adult immunization is an area of prevention with many opportunities to improve population health. This is particularly true for at-risk adults with health conditions that place them at higher risk of developing complexities that can lead to costly and serious health complications.

In response to this serious public health problem, AMGA conducted 3, one-year learning collaboratives, the Adult Immunization Best Practices Learning Collaboratives (“the Collaboratives”), to address barriers, share best practices, and improve pneumococcal and influenza vaccination rates. The results of the first pilot collaborative (Cohort 1) are reported elsewhere. In an effort to scale up and spread improvements realized in the pilot program, 2 subsequent collaboratives (Cohorts 2 and 3) leveraged learnings from pilot Cohort 1, which included a formal qualitative component that identified organizational, individual, cultural, and contextual factors that influenced the success of the programs. The current study examined the ability to scale up and spread the findings from the 7-site pilot program (Cohort 1) to 39 additional health care organizations that participated in 2 subsequent collaboratives (Cohorts 2 and 3) by applying IHI’s Framework for Spread. Vaccination rates were assessed through a rigorous quantitative analysis using an approach that enabled true effects to be estimated by comparing providers at participating organizations with providers at other organizations with similar baseline characteristics and therefore a similar propensity to join a learning collaborative. This matching technique prevented overestimation of the impact of the intervention. An informal qualitative analysis helped support the quantitative data and validated the findings from the pilot study.

Methods

Ethics approval and consent to participate

The study was determined exempt by the New England Independent Review Board (NEIRB# WO 1-776-1) for pilot study. Further information and documentation of Institutional Review Board exemption is available upon request.

Study sample

The study population included adult patients who received health care services between 2017 and 2018 at 9 large US health care organizations in 7 states, including independent medical groups and integrated delivery systems, that participated in the 2 Collaboratives, described in the Intervention section, and patients from 29 matched-control organizations. These 9 health care organizations are a subset of the 39 organizations that participated in the Collaboratives and were selected based on patient- and provider-level data availability in the Optum® database. Study sites used a shared Optum population health analytics platform and contributed data to the common data repository (CDR) (described in the Data Source section). Table 1 provides descriptive information about each cohort. The 9 organizations included in the analysis reflected the characteristics of the 30 remaining organizations in terms of size, structure (eg, integration), and baseline immunization rates. In other words, the 9 study organizations were representative of the remaining 30. Three organizations participated in both Cohort 1 and either Cohort 2 or 3.

Data source

AMGA is a nonprofit trade association representing 420 multispecialty medical groups and integrated health care delivery systems with a total of 175,000 full-time equivalent physicians. As AMGA’s distinguished data and analytics collaborator, Optum provides access to data from AMGA members, many of whom use Optum Analytics’ population health tools. In addition to claims data, Optum Analytics tools include clinical data from members’ electronic health records (EHRs), mapped and normalized to allow valid, reliable comparisons. Detailed EHR data enable discovery of differences in care process as well as clinical outcomes. The CDR that pools longitudinal EHR data from 54 health care organizations, including records for approximately 79 million patients, was accessed for this study. Vaccination data were pulled from several areas in the EHR including medication tables, medication patient reports, vaccination tables, health maintenance tables, procedure codes, and diagnosis codes.

Study design

In order to assess the success of the scale-up and spread of a pilot learning collaborative (Cohort 1), vaccination data were examined to identify changes in adult vaccination rates over time. Provider-level propensity scores were used to match intervention to nonintervention providers to control for inherent selection bias of participating organizations. Based on a review of existing literature and subject matter expertise, 6 variables chosen for the propensity score models included baseline patient count, baseline vaccination rate(s), baseline minority patient rate, baseline Medicare
patient rate, baseline Medicaid patient rate, and baseline Medicare wellness visit rate. Comparative analyses were conducted between intervention sites and nonintervention sites on vaccination rates using a difference-in-differences approach. Table 1 includes provider baseline rates by cohort. Qualitative data collected during site visits and in-person and virtual meetings helped identify system- and clinic-level interventions, and organizational factors contributing to successful programs.

Lists of participating providers were obtained from the 9 collaborative health systems. In order to be included in each provider’s patient population for a given measure, the patient needed to satisfy the age restriction for the measure (age at start of time period), have a qualifying ambulatory visit during the time period, be alive for the full time period, and be documented in the EHR data as having a designated primary care provider (PCP) for the most days during the 24 months leading up to the end of the time period. For the pneumococcal high-risk and pneumococcal at-risk measures, the patient also needed to have a diagnosis for one of the high- or at-risk conditions between January 2013 and the end date of each intervention period. Table 2 displays specifications for each measure.

Matching variable statistics were obtained for the aforementioned 6 variables used for propensity score matching. Provider-level vaccination rates were calculated for each measure. For the pneumococcal 65+ and pneumococcal high-risk measures, vaccination rates of interest included “ANY” (patients with any vaccination, including records of pneumococcal polysaccharide vaccine [PPSV] administration, PCV administration, or administration of unknown pneumococcal vaccine) and “BOTH” (patients with documentation of receiving both PPSV and PCV vaccinations). For the pneumococcal at-risk measure, vaccination rates of interest included “ANY” (patients with documented PPSV or administration of unknown pneumococcal vaccine) and “PPSV” (patients with documented PPSV). In the case of the pneumococcal 65+ measure, vaccinations were only considered if they occurred on or after the patient’s 65th birthday. For the pneumococcal high- and at-risk measures, vaccinations were only considered if they occurred during ages 19 through 64. For the influenza measure, vaccinations had to occur within the current flu season (ie, July 1 through June 30), and on or after a patient’s 18th birthday to be considered.

Minority rate was defined as the proportion of nonwhite patients; patients with unknown race were classified as white and patients reporting multiple races were classified as nonwhite. Medicare and Medicaid rates were defined as the proportion of patients who had at least 1 insurance record of the given product type or at least 1 month of eligibility (where claims data were available) during the baseline period. Patients with dual membership for both Medicare and Medicaid products were included in the numerator for both rates. Medicare wellness visit rate was defined as the proportion of patients aged ≥65 years who had at least 1 Medicare wellness visit (Healthcare Common Procedure Coding System codes G0402, G0439, and G0438) during the baseline period.

The comparison provider population was created using 29 health systems and organizations available in the database, but not participating in the adult vaccination learning collaborative. Data on the 6 matching variables were collected for all providers who saw at least 50 patients during both the baseline and intervention periods.

**Propensity score matching**

The propensity score was estimated using a logistic regression model, in which treatment status (collaborative participant or not) was regressed on the aforementioned 6 observed baseline characteristics. The estimated propensity score is the predicted probability of participating in the collaborative, derived from the fitted regression model. Participating collaborative providers (intervention) were one-to-one matched to non-collaborative providers (control), with replacement, on the logit of the propensity score. All

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### Table 1. Collaborative Cohort Descriptions

<table>
<thead>
<tr>
<th></th>
<th>Cohort 2</th>
<th>Cohort 3</th>
</tr>
</thead>
<tbody>
<tr>
<td>Collaborative Intervention Periods</td>
<td>1/2017–3/2018&lt;sup&gt;2&lt;/sup&gt;</td>
<td>7/2017–6/2018</td>
</tr>
<tr>
<td>Sites, n</td>
<td>5</td>
<td>4</td>
</tr>
<tr>
<td>Providers, n</td>
<td>414</td>
<td>152</td>
</tr>
<tr>
<td>Control sites used, n</td>
<td>21</td>
<td>27</td>
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<tr>
<td>Control sites considered</td>
<td>23</td>
<td>27</td>
</tr>
<tr>
<td>Integrated Systems, %</td>
<td>94</td>
<td>53</td>
</tr>
<tr>
<td>Eligible patients, pneumococcal, n</td>
<td>~541,000</td>
<td></td>
</tr>
<tr>
<td>Eligible patients, influenza, n</td>
<td>~858,000</td>
<td></td>
</tr>
</tbody>
</table>

1. For illustrative purposes, these numbers reflect the pneumococcal vaccination measure for adults ≥65 years. The rates differ slightly for each of the 4 quality measures.
2. For cohort 2, influenza baseline and intervention time periods were adjusted to contain measurements within a single flu season.

PCV, pneumococcal conjugate vaccine; PPSV, pneumococcal polysaccharide vaccine; SD, standard deviation.
providers were matched. Matching without replacement and/or with a caliper also were explored. To enable a sub-
group analysis comparing integrated delivery system (IDS) providers with ambulatory providers, a block-randomized de-
sign was mimicked by propensity score matching within these 2 subgroups (IDS providers were matched to IDS
providers and ambulatory providers to ambulatory providers). To allow for effect estimates at the health system level, in-
dividual propensity score models were created for each health
system so that the model would predict the best matches for
providers within each individual health system. Because the
ultimate goal of propensity score models is balance—not
satisfaction of model diagnostics—transformations and in-
teraction terms were explored. Using these transformations and higher order terms, several propensity score models
were created per health system.

For each propensity score model, 500 iterations were
tested, producing different matches with each iteration. The
different matches were prompted by using a “distance tol-
erance,” in which all control providers within a certain
distance are treated as ties, and one provider is randomly
selected from the tying set. This allows for greater fluctua-
tion when matching with replacement. In the end, more than
5000 distinct matched sets were produced per health system.
To evaluate the quality of the matched sets, standardized
differences were calculated. The sum of the standardized
differences across all covariates was calculated to select the
best matched set for each health system. The best matched
sets for each health system were aggregated to evaluate the
sum standardized difference overall, within the IDS and am-
bulatory subgroups, and at the individual health system level.

Other matching diagnostics included constructing quantile-
quantile plots, overlaying histograms, and overlaying dens-
ity plots to compare the distribution of continuous baseline
covariates between intervention and control providers.
Means and standardized differences of each covariate were
computed for the intervention and control provider groups.
These balance diagnostics were evaluated overall and within
the IDS and ambulatory subgroups. The number of times
each control was used overall and for each individual health
system also were monitored.

Using the best matched set, generalized estimating equa-
tions (GEEs) were applied to compute a difference-in-differences estimate for each vaccination rate of interest,
controlling for the inherent correlation between a provider’s
baseline and intervention rates. As a part of the GEE, var-
iances were adjusted to account for some controls being reused (because of matching with replacement). The re-
ported estimates represent the improvement of intervention
providers above that of their matched controls.

Finally, sensitivity of the reported estimates to a partic-
ular matched set was analyzed using the best 500 matched
sets (of the total ≥5000 matched sets). These matched sets
have nearly as good, though not the best, matching diag-
nostics. Presumably, unmeasured confounders are shifting
around in these distinct matched sets, so significant differ-
ences demonstrated in all 500 matched sets is indicative of
robustness to unmeasured confounders.

**Qualitative analysis**

Lessons learned from the formal qualitative analysis col-
clected during the pilot program11 were leveraged to inform
collaborative Cohorts 2 and 3 (current study). Interventions
and self-identified best practices were recorded during Cohort
1 and organized into a framework that was provided to Co-
horts 2 and 3. This allowed for learnings from Cohort 1 to be
transferred to Cohorts 2 and 3. The framework grouped in-
terventions into domains and provided an estimated level of
difficulty of each intervention to accommodate the varying
levels of resources and implementation readiness of each
organization. As in other learning collaboratives, a non-
prescriptive approach was taken; interventions other than
those in the framework could be implemented. In addition, in
this study informal qualitative data were collected during site
visits and during both in-person and virtual meetings. These
data were collected to confirm findings from the pilot study
and to discover potential new strategies used during Cohorts
2 and 3. Barriers, facilitators, and most successful strategies
(as perceived by participants) were identified.

**Framework for spread**

The IHI created a “Framework for Spread”14 more than
a decade ago that serves as a useful guide to identifying
key components that contribute to effective scale-up and
spread of new ideas, innovations, and operational systems
both within and across organizations. Although collabora-
tives themselves have proved successful interventions12,13 in
scale-up and spread, the Framework for Spread helps hone
in on the specific, necessary components, including leader-
ship support, identification of better ideas, communication,
strengthening the social system, measurement and feedback,
and knowledge management. The authors recognize that
organizational context plays an important role and factors
such as culture, infrastructure, size, strength of social sys-
tem, and the intervention or system being spread influence
how components of the framework are applied.

**Intervention**

The study intervention was the learning collaborative
approach taken to improve adult vaccination rates over

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**Table 2. Measure Specifications**

<table>
<thead>
<tr>
<th>Measure</th>
<th>Eligible ages</th>
<th>Eligible conditions</th>
</tr>
</thead>
<tbody>
<tr>
<td>1: Pneumococcal ≥65</td>
<td>≥65</td>
<td>All cerebrospinal fluid leak, cochlear implants, hemoglobinopathies, asplenia, chronic renal failure, nephrotic syndrome, organ transplant, kidney disease, immunodeficiencies, leukemia, lymphoma, Hodgkin's disease, multiple myeloma, general malignancy</td>
</tr>
<tr>
<td>2: Pneumococcal High Risk</td>
<td>19–64</td>
<td>All diabetes, chronic heart conditions, lung disease, chronic liver disease, alcoholism, smoker</td>
</tr>
<tr>
<td>2a: Pneumococcal At Risk</td>
<td>19–64</td>
<td>All diabetes, chronic heart conditions, lung disease, chronic liver disease, alcoholism, smoker</td>
</tr>
<tr>
<td>3: Influenza</td>
<td>≥18</td>
<td>All</td>
</tr>
</tbody>
</table>
3 one-year collaborative periods and is reported in detail elsewhere. Briefly, the learning collaboratives were led by an expert advisory committee and included in-person meetings, webinars, sharing of best practices, education, site visits, goal setting, outreach and coaching, peer-to-peer learning, case studies, and clinical outcome measurement. Participating organizations were encouraged to choose from a selection of strategies to improve their vaccination rates including, but not limited to, nurse standing orders, use of a 1-way or 2-way state vaccination information system registry, patient and provider education, patient outreach, EHR registries, health maintenance and best practice alerts, working with specialists, working with pharmacies, designating provider champions, and identifying and learning from high performers. Participating organizations chose a variety of strategies. Learnings from the pilot study were heavily leveraged in Cohorts 2 and 3 in the spirit of progressive and iterative learning. Although a total of 39 organizations participated in either Cohort 2 or 3, only 9 were included in this study because of their data availability in the aforementioned Optum data set.

Results

Quantitative

Among the 9 study organizations, there were 250k, 99k, 192k, and 858k eligible patients for pneumococcal ages ≥65 years, pneumococcal high risk, pneumococcal at risk, and influenza vaccines, respectively. Table 3 shows baseline unadjusted (ie, unmatched) and adjusted (ie, matched) characteristics of the intervention and comparison cohorts for PV for patients aged ≥65 group for Cohort 2. (Cohort 3 characteristics are similar; data available upon request.) For the pneumococcal ≥65 measure, a total of 414 collaborative PCPs were matched to 268 non-collaborative providers out of a pool of 4290 providers. The sum standardized difference (SSD) for this measure and cohort was reduced from 5.46 to 0.309; similar reductions were observed across other measures and cohorts, indicating well-matched study samples. Across 4 measures in Cohorts 2 and 3, the overall adjusted SSDs ranged from 0.113 to 0.353.

Table 4 shows the average treatment effect of the intervention for each vaccination and type of vaccination overall for each cohort. In 2017–2018, both intervention cohorts demonstrated greater improvement than their matched providers in PVs (both vaccines PPSV and PCV) for patients age ≥65 (treatment effect range: 1.4%-3.7%, P<0.01) and PV for high-risk patients (eg, with immunocompromising conditions) aged 19–64 (0.8%-1.6%, P<0.01). Significant effects were observed in Cohort 3 for PV for at-risk patients (eg, with diabetes) aged 19–64 (1.7%, P<0.01), and influenza vaccination rates (2.4%, P<0.001). Individual health systems demonstrated even greater improvements across all 4 vaccinations: 9.5% influenza; 8.7% PV age ≥65; 11.8% PV high risk; 16.3% PV at risk (all P<0.01). Table 5 presents pre–post vaccination rates for intervention and matched control groups across immunization type and cohort, for adult vaccinations recommended by the CDC. Qualitative

Although formal qualitative interviews were not conducted for Cohorts 2 and 3, site visits were conducted at 20 of the 39 health care organizations. During the site visits, AMGA staff members informally discussed progress on sites’ individual adult immunization programs. Site visit qualitative findings were supplemented with information gathered during 2 in-person meetings, 12 webinars, and regular conference calls.

Commonly reported facilitators of adult immunization were similar to the pilot cohort and included provider education, implementing best practice alerts in the EHR, and collaboration with specialists, particularly for patients who were at risk or high risk for pneumococcal disease, and the collaborative itself. When asked what they liked best about the collaboratives, participant comments included “coming together with multiple organizations around the country tackling the same issue, networking very enjoyable, hearing best practices, collaboration, great discussions, ideas shared, great way to connect with peers.” Engagement of pharmacists to administer immunizations was noted more often as a facilitator in Cohorts 2 and 3. This was attributed to additional education about the role pharmacists can play in increasing immunization rates provided during Cohorts 2 and 3. Additional facilitators were the presence of immunization “champions” within the organization, automated patient outreach, and improving the exchange of data with state immunization registries.

A commonly reported barrier was the complexity of the immunization guidelines for patients with high-risk and at-risk conditions. Documentation issues were a lesser barrier for Cohorts 2 and 3 than they were for the pilot cohort. This was attributed to the addition over time of structured fields in EHRs to reflect changes in adult pneumococcal vaccine guidelines.

The most influential factors in successful spread align with IHI’s Framework for Spread, including the importance of leadership buy-in and support, the identification of better ideas or practices than what was currently being done, measurement and feedback to providers and other key program staff, and knowledge management in the form of education for all staff members, especially around PVs for high- and at-risk patients.

Discussion

The adult immunization best practices learning collaborative was successfully scaled from 7 pilot sites (Cohort 1) to 39 expansion sites (Cohorts 2 and 3). Results demonstrated that expansion cohorts performed as well or better than the pilot sites. For example, expansion cohorts showed greater treatment effects for the recommended vaccinations for PV (eg, PPSV and PCV for PV age ≥65 and PV high risk) versus any vaccine (ie, PPSV or PCV). This suggests there is progress toward patients receiving the recommended vaccines as there are time restrictions preventing meeting requirements immediately (eg, recommended administration of PPSV is 1 year following receipt of PCV). Only Cohort 3 showed a positive treatment effect for influenza vaccination. This might reflect better documentation of vaccinations received outside the clinic environment. This was a noted primary barrier in the pilot cohort and therefore received significant focus during the expansion collaboratives. Improvements to state immunization registries may have contributed to better documentation.
An impressive finding also demonstrated during the pilot collaborative (Cohort 1), among high- and at-risk populations was that PV improvement rates increased following the difference-in-difference analysis because of decreasing immunization rates among control group providers. For example, at one site rates for at-risk patients increased from 52% to 65% while their matched control providers demonstrated a rate decrease from 55% to 52%. Again, the high- and at-risk populations were a greater challenge and less attention has been paid to these groups nationally.

The Collaborative itself was an important intervention leading to successful scale-up and spread, as were the key components identified during the pilot (Cohort 1) and emphasized during the expansion collaboratives that aligned with IHI’s Framework for Spread. While participating in the Collaborative, and even upon conclusion, health care organization representatives learned and borrowed ideas from each other, troubleshooting problems, and received guidance from national experts. The key components emphasized across collaboratives included leadership support and buy-in, measurement and feedback, communication, testing of innovations, and education or knowledge management. These align with the successful components identified and supported by the Framework for Spread.

As with the pilot study, a rigorous analytic approach was taken to compare the intervention organizations with other...
motivated and committed organizations. Typically, quality improvement programs are evaluated using a pretest–posttest design, often overestimating treatment effects and, ultimately, the impact of interventions. The approach taken in this study enabled true effects to be estimated by comparing providers at participating organizations with providers at other organizations with similar baseline characteristics and therefore a similar propensity to join a learning collabora-
tive. This matching technique prevented the overestimation of the impact of a learning collaborative approach. Indeed, had participating organizations been compared to all health systems in the United States, a much greater difference would have been observed. Therefore, the authors can conclude that participating systems, even when compared to the most high-performing health systems available, still improved at least twice as much.

Table 5. Adjusted Pre–Post Intervention Immunization Rates Compared to Matched Controls

<table>
<thead>
<tr>
<th>Immunization</th>
<th>Intervention</th>
<th>Matched controls</th>
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</tr>
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<tbody>
<tr>
<td>Pneumococcal At Risk: Cohort 2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pneumococcal High Risk: Cohort 2</td>
<td></td>
<td></td>
<td></td>
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<tr>
<td>Influenza</td>
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1Difference-in-difference analysis comparing intervention to matched controls.
PCV, pneumococcal conjugate vaccine; PPSV, pneumococcal polysaccharide vaccine.

Limitations

There are limitations to the quantitative analyses. Three health care organizations that participated in Cohort 1 also participated in either Cohort 2 or 3. Although having had a longer overall time to make improvements may have fa-
vored their improvement, it also could have been more difficult for them to continue to improve on some measures. The limited pool of high-performing providers available as control providers meant the authors needed to match with replacements, which caused variance estimates to increase, and some control providers potentially had a greater influence on the reported effect estimates. In addition, there only were data on 9 of 30 organizations for the comparison analysis. However, the authors were able to confirm that the 9 study groups were representative of the larger group of 30 in terms of organizational size, structure, and baseline immu-

nization rates. Although qualitative data collected during site visits and meetings supported findings from the qualitative data collection in the pilot study, formalized inter-
views conducted in Cohorts 2 and 3 may have enriched the analysis. A cost analysis of this approach was not conducted. The focus was on clinical population health outcomes and methods to improve these outcomes at a population level. It also was beyond the scope of the study to follow vaccinated patients to determine whether they experienced improved outcomes in terms of fewer hospitalizations or lower over-
all costs. Finally, the authors did not formally examine race/ethnic disparities in adult immunization rates. How-
ever, one participating organization did focus on disparities and was able to improve influenza vaccination rates for
Hispanic and African American patients, decreasing the overall gap in rates.

**Conclusion**

In summary, these findings suggest a successful scale-up and spread of an intervention tested as a pilot study. These results further support a learning collaborative approach to improving population health outcomes in the area of adult vaccinations, especially for vaccinations with lower baseline national rates. In an era when health care dollars are scarce, leaders need to know with a degree of certainty that the resources they deploy to improve population health will result in desired outcomes. This study demonstrates how a learning collaborative approach can be successfully piloted and subsequently scaled up and spread. But perhaps even more importantly, it demonstrates how to identify those strategies and interventions that work best in real world, clinic, and organizational settings.

**Acknowledgments**

The authors wish to acknowledge the 39 AMGA member health care organizations that participated in the 2 learning collaboratives described in this manuscript, as well as the 7 organizations that “paved the way” by participating in the pilot study. Their commitment to improving adult immunization rates made this study possible. We also acknowledge the Population Health Initiatives Team at AMGA Foundation, without whom the best practices learning collaboratives would not have been possible. Their dedication, skill, and experience in conducting best practices learning collaboratives make work like this possible.

**Author Disclosure Statement**

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**References**


Address correspondence to:

Elizabeth L. Ciemins, PhD, MPH, MA
AMGA Analytics
One Prince Street
Alexandria, VA 22314-3318
USA

E-mail: ecиемins@amga.org