

Design and implementation of a randomized field experiment measuring benefits of medical scribes in the VA

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Conflicts of Interest

None to declare.

ABSTRACT

Background

Medical scribes are trained professionals who assist health care providers by administratively expediting patient encounters. Section 507 of the MISSION Act of 2018 mandated a two-year study of medical scribes in VA Medical Centers. This study began in 2020 in the emergency departments and specialty clinics of twelve randomly selected VA Medical Centers across the country, in which 48 scribes are being deployed.

Methods

We are using a two-arm randomized field experiment to assess the effects of medical scribes on productivity, wait times, and patient satisfaction in selected specialties in the VA that traditionally have high wait times. Scribes will be assigned to emergency departments and/or specialty clinics (cardiology, orthopedics) in the VA Medical Centers randomized into treatment. Remaining sites that expressed interest in the study but were not randomized to treatment will be used as a comparison group.

Results

Our early implementation findings indicate that contracting seems to hold a significant advantage over VA employment in terms of reaching hiring targets but took somewhat longer from job posting to start date.

Conclusions

Our evaluation findings will provide insight into whether scribes can increase provider productivity and decrease wait times for high demand specialties in the VA without adversely affecting patient satisfaction.

Implications

As a learning health care system, this trial has great potential to increase our understanding of the potential effects of scribes while also informing a real policy problem in high wait times and provider administrative burdens.

Key Words

Veterans, MISSION Act, medical scribes, productivity, wait times

MAIN TEXT

Introduction

Section 507 of the VA MISSION Act of 2018, formally the John S. McCain III, Daniel K. Akaka, and Samuel R. Johnson VA Maintaining Internal Systems and Strengthening Integrated Outside Networks Act of 2018, mandated a two-year pilot study of the introduction of medical scribes in VA Medical Centers (VAMC), focused on emergency departments (ED) and high wait time specialty clinics.¹ Medical scribes are trained professionals but not clinically licensed, assisting health care providers by administratively expediting an episode of care. Prior to this study, scribes have rarely been employed by the Veterans Health Administration (VHA). According to the law, scribes will assist providers in navigating and documenting patient information into the electronic health record (Appendix A). Our objective was to build a randomized evaluation around the requirements of the pilot to better understand how the introduction of scribes affects provider productivity and patient satisfaction.

Outside of the VHA, evidence suggests that scribes may increase provider productivity and satisfaction, and decrease provider time spent on documentation without affecting patient satisfaction.^{2,3} A randomized trial in a family medicine clinic in Northern California found no effect on patient satisfaction and significant improvements in physician satisfaction and productivity.⁴ In a study where seven scribes were introduced to an urban safety net primary care clinic, face-to-face time increased by 57% and computer time associated with visits decreased by 39%.⁵ There were no changes in total visit or provider cycle time, but work relative value units (wRVUs) and patients per hour increased. Patients were less likely to report being “very comfortable” with the number of people in the exam room following the introduction of scribes. Another recent study assessed the effect of scribes in a suburban, non-academic, community ED,

finding consistent declines in wait times, visit times, and total length of stay coupled with improvements in patient and provider satisfaction.⁶ Similarly, wRVUs and patients per hour increased with chart review and post-visit documentation time falling. Studies in a tertiary academic ED corroborated others' findings of decreased provider documentation burden and an increase in RVUs for adult patients with mixed findings for length of stay.⁷⁻¹⁰ Outside of the US, a multicenter randomized trial in five Australian EDs found a 16% increase in physician productivity and a 19 minute reduction in median length of stay.¹¹ However, none of this evidence was generated in a VA setting and was generally on a much smaller scale than this congressionally mandated study.¹² Process, staffing, and patient mix differences could yield different results and several of the studies used an uncontrolled pre-post design, limiting causal interpretation of their results.

This two-year study officially began on June 30, 2020 in the EDs and specialty clinics of twelve VAMCs across the country, in which 48 scribes are being deployed. The goal of this evaluation is to understand how the introduction of scribes affects provider efficiency, wait times, daily patient volume, and patient satisfaction.

Methods

Setting and randomization

Section 507 of the MISSION Act specified that at least four participating medical centers must be located in urban and rural areas along with two in underserved areas with a need for increased access.¹ OVAC developed an initial list of 32 VAMCs interested in participating in the study, which were then categorized based on rurality and location in an underserved area. OVAC is actively exploring additional metrics to measure access to care beyond wait times but for the purposes of this study, underserved status was determined based on high specialty care wait times

for new patients following congressionally-defined standards.¹⁴ VAMCs were sorted into categories by location (urban, rural, underserved) and specialty (ED, specialty care), according to the requirements of the MISSION Act, leadership preferences, and site capabilities before randomization. To represent specialty care while minimizing heterogeneity in the study, leadership chose to focus on cardiology and orthopedics. VAMCs were then randomly assigned to the treatment group within each location-specialty, which yielded seven urban (two of which are underserved) and five rural VAMCs assigned to treatment.

The MISSION Act also specified that 30% of the scribes were to be assigned to emergency departments and the other 70% to specialty care. New Jersey requested to have their scribes split between the emergency department and specialty care. Two other sites (Clarksburg, West Virginia and Hampton, Virginia) requested to split their scribes within specialty, across cardiology and orthopedics. The randomization and these requests should yield 31 scribes in specialty care (65%) and 17 in emergency departments (35%), which is close to the targeted split in the Act. Five backup sites were also selected should one or more chosen sites fail to participate, based on being next in the randomization order for a given location-specialty combination. The remainder of the VAMCs that expressed interest in the study are being used as a comparison group, including any backup sites that do not transition into treatment during the study. The mix of locations and specialties randomly assigned to implement scribes as well as the backup and comparison sites are shown in Table 1.

Table 1. Study locations and specialties

Location	Specialty	Randomization group	Station name (number)
Urban	Emergency department	Treatment	New Jersey (561) ^a Temple, TX (674) Southern Arizona (678) San Antonio, TX (671)
		Backup	Indianapolis, IN (583)
	Comparison	St. Louis, MO (657) Reno, NV (661) Las Vegas (654) Salt Lake City, UT (660) Minneapolis, MN (618) Salisbury, NC (659) Greater Los Angeles, CA (603) Louisville, KY (603) Erie, PA (562) Hampton, VA (590)	
		Treatment	New Jersey (561) ^a Louisville, KY (603)
		Backup	Northport, NY (632)
		Comparison	Long Beach, CA (600) St. Louis, MO (657) Minneapolis, MN (618) Southern Arizona (678) Greater Los Angeles, CA (603) Salt Lake City, UT (660) Cincinnati, OH (539) Salisbury, NC (659)
Rural	Emergency department	Treatment	Togus, ME (402)
		Backup	St. Cloud, MN (656)
		Comparison	Clarksburg, WV (540) New Mexico (501) Fargo, ND (437) Salem, VA (658)
	Specialty care	Treatment	Montana (436) Fargo, ND (437) Clarksburg, WV (540) Manchester, NH (608)
		Backup	Salem, VA (658)
		Comparison	St. Cloud, MN (656)
Underserved	Specialty care	Treatment	Hampton, VA (590) Oklahoma City, OK (635)
		Backup	Columbus, OH (757)
		Comparison	Durham, NC (558) Montana (436) Asheville, NC (637)

Backup sites will be used as comparison sites unless a treatment site fails to hire and implement scribes.

^a After randomization, New Jersey requested to split scribes between their emergency and specialty care departments.

Intervention

The randomization of VAMCs was finalized in April 2019 with hiring of scribes beginning in November 2019 with the two-year study period officially beginning June 30th, 2020, having been delayed several months by the onset of the COVID-19 pandemic. Four medical scribes are to be assigned to each of the VAMC sites randomized into treatment with the VA hiring half as new employees and the remaining half as contractors. Two scribes, ideally one VA employee and one contractor, will be assigned to each participating provider, with two providers participating in the study at each facility. The VHA is relying on providers volunteering for pairing with a scribe, which limits generalizability as it has in other studies.¹³ Adoption of scribes is voluntary in this study but likely would not be in a broader implementation, which has implications for scaling the effects identified. We are also interested in whether the mode of hiring, as VA employees or as contractors, plays any role in scribes' effectiveness.

The goal is to keep the provider-scribe pairs consistent throughout the study to the extent possible. Scribes will work with other Licensed Independent Practitioners (LIPs) in their assigned specialty if their provider partner is not available. Pairing two scribes with each provider allows the provider to rotate between scribes for face-to-face patient encounters while the other scribe has time to finish their notes on the prior patient. Clinical notes taken by scribes must be tagged by each scribe with their name, date, and time and approved by the provider before becoming viewable in the electronic health record. This will also allow us to identify which visits involved scribes for the purposes of evaluating the program. We have included a program training manual and policy statement describing how scribes will assist in documentation of patient encounters (Appendix B and C).

Study design

We are using a two-arm randomized field experiment to assess the effects of medical scribes on productivity, wait times, and patient satisfaction, incorporating both within-facility controls where appropriate (e.g., more difficulty in the ED setting as providers are floating from patient to patient) and comparing across treatment and control sites. Using both within- and between-facility variation is important as scribes may not only affect productivity for their assigned provider, but have spillover effects elsewhere in the clinic. For providers in specialty clinics, the use of scribes will be primarily focused on outpatient visits; however, they may also round on the inpatient service as needed or desired. Provider productivity, patient volume, wait times, and patient satisfaction from the treated sites, using the measures noted below, will be compared to baseline (pre-intervention) data as well as data from comparison sites. We will note the impact of scribes in different clinical settings, including rural versus urban, specialty clinic versus ED, and underserved areas. A schedule of enrollment, interventions, and assessments is shown as Table 2. This study is being coordinated by the Partnered Evidence-based Policy Resource Center (PEPREc) at the VA Boston Healthcare System in collaboration with the VA Office of Veterans Access to Care (OVAC), the VA Collaborative Evaluation Center (VACE), and participating VAMCs across the US.

Table 2. Schedule of enrollment, interventions, and assessments

TIMEPOINT	STUDY PERIOD				
	Enrollment	Allocation	Post-allocation		
	March 2019	April 2019	July 2020	2021	June 2022
ENROLLMENT:					
Eligibility assessment for interested facilities	X				
Randomization		X			
INTERVENTIONS:					
<i>Medical scribes</i>			←—————→		
<i>No medical scribes</i>					
ASSESSMENTS:					
<i>Personnel records</i>		X	X	X	X
<i>Patient visit records</i>		X	X	X	X
<i>Patient satisfaction surveys</i>		X	X	X	X

As implementation of the scribes pilot program was congressionally mandated, it is part of VA’s required operations and therefore exempt from Institutional Review Board review. A memorandum from OVAC (Appendix D) directs PEPReC to design and execute this evaluation using secondary data that is already collected by VHA in its routine clinical operations and human resources functions. The study was registered on ClinicalTrials.gov (NCT04154462) in November 2019 as hiring of scribes began (Appendix E). The only deviation from our ClinicalTrials.gov record is that hiring had been expected to be completed by March 2020, allowing the study implementation to begin, but the COVID-19 pandemic resulted in a several month delay.

Data sources

We will use several data sources to develop the outcome measures described in Table 3 with detailed descriptions contained in Appendix F. We will use appointment, visit, and procedure data from the VHA Corporate Data Warehouse (CDW) to capture wait times for patients, visits, and services performed by physicians. We will use the Personnel and Accounting Integrated Data (PAID) database from the VHA Workforce Management and Consulting Office to obtain data on work hours during the study period for capturing full-time equivalents (FTE) for physicians and providers in each pay period included in the study. Patient satisfaction will be assessed using survey data from Veterans Signals (V-Signals), a nationally standardized tool aimed at better understanding the Veteran experience and satisfaction with the care received at VAMCs. V-Signals is an ongoing email-based survey conducted by the Veterans Experience Office (VEO), with over 3 million lifetime responses and a historical response rate around 20%. Any Veteran receiving outpatient care within the last week is eligible to receive a survey and the survey remains open for two weeks after invitation. These surveys will also be supplemented in a qualitative evaluation of the pilot being conducted by VACE. We are also using data from the VHA Planning Systems Support Group (PSSG) and Support Service Center (VSSC) databases, American Community Survey, Centers for Medicare and Medicaid Services, and Zillow to develop measures of treatment and control variables summarized in Table 4 and described in detail in Appendix F.

Table 3. Outcome measures

Domain	Measure	Level
Provider efficiency	Work relative value-based provider efficiency	Facility-pay period, provider-pay period
	Visit-based provider efficiency	Facility-pay period, provider-pay period
	Daily visit-based provider efficiency	Facility-pay period, provider-pay period; scaled by FTE days
Wait times	Days to completed consult	Facility-pay period, provider-pay period
	Days to scheduled consult	Facility-pay period, provider-pay period
Patient volume	Unique patient volume per day	Facility-pay period, provider-pay period
Patient satisfaction	“It was easy to get my appointment”	Facility-pay period
	“After I checked in for my appointment, I knew what to expect”	Facility-pay period
	“I got my appointment on a date/time that worked for me”	Facility-pay period
	“I trust this clinic for my healthcare needs”	Facility-pay period
	“My provider listened carefully to me”	Facility-pay period
	“My provider explained things in a way that I could understand”	Facility-pay period
	“I am satisfied with the service I received from the VA clinic”	Facility-pay period

Table 4. Treatment measures and control variables

Measure	Level	Source
<i>Treatment measures</i>		
Scribe FTEs per 1,000 patients (<i>provider efficiency, wait times, and patient volume models</i>)	Facility-pay period	Personnel and Accounting Integrated Data
Physician FTEs per 1,000 patients (<i>wait times models</i>)	Facility-pay period	Personnel and Accounting Integrated Data
Quartiles of scribe FTEs per physician FTE (<i>patient satisfaction models</i>)	Facility-pay period	Personnel and Accounting Integrated Data
<i>Control variables</i>		
Percentage of enrollees over age 65	Facility-year, derived from county-level	Planning Systems Support Group
Percentage of enrollees under age 50	Facility-year, derived from county-level	Planning Systems Support Group
Percentage of high priority status enrollees (7 and 8)	Facility-year, derived from county-level	Planning Systems Support Group
Insured rate for 18- to 64-year- old males	Facility-year, derived from county-level	American Community Survey
Median household income	Facility-year, derived from county-level	American Community Survey
Veteran unemployment rate	Facility-year, derived from county-level	American Community Survey
Home prices	Facility-year, derived from county-level	Zillow Home Value Index
Medicare Advantage (MA) penetration	Facility-year, derived from county-level	Centers for Medicare and Medicaid Services
Average patient risk scores	Facility-year	Support Service Center
Average enrollee driving distance ^a	Facility-year	Planning Systems Support Group
Average community care wait times for specialty care ^a	Facility-year	Corporate Data Warehouse

^a Applies to specialty care only

Statistical analysis

Our analysis of early implementation will focus on the trajectory of hiring of VA versus contract scribes and average time to hire. These data will be based on the ‘Entrance on Duty’ (EOD) date, when the scribe has completed all the necessary steps to begin working. We will describe achieved and projected hiring trends, including scribes that are currently going through

the onboarding process but have not yet reached their EOD date. Their projected EOD dates were calculated by using the most conservative estimate of onboarding time available (the scribe that took the longest to reach their EOD date), adding that time to the date when a scribe began onboarding.

For our outcome evaluation, we will measure the impact of medical scribes on provider efficiency, wait times, patient volume, and patient satisfaction, using variation between facilities and providers having and not having scribes as well as pre-intervention baseline outcome data across treatment and comparison sites. As mandated by law, we will also study differences in provider efficiency among providers with VA scribes and those with contracted scribes to the extent possible. Our ability to assess the latter will be dependent on how VA and contracted scribes ultimately are distributed across facilities and physicians.

We will use descriptive statistics and multivariable regression analysis to describe the impact of scribes on the outcome measures summarized in Table 3. We plan to use facility and provider-level fixed effects models that exploit within-facility and within-provider variation in outcomes and presence of scribes over time, incorporating the control variables described below and summarized in Table 4. We will also account for supply and demand factors for VA health care known to be associated with our outcomes,¹⁵ assuming there is enough variation over time to justify their inclusion in addition to facility fixed effects. These include facility-level enrollment characteristics, such as percentage of enrollees over age 65, percentage under age 50, and percentage of low priority status enrollees (7 and 8) from PSSG. PSSG captures enrollee counts at the county level for each facility, which we aggregate to the appropriate facility level. We also include the facility-level insured rate for 18- to 64-year-old males (proxy for veteran insurance coverage), median household income, home prices, veteran unemployment rate, and Medicare

Advantage penetration, derived from county-level measures and weighted by enrollment in each VA facility. Facility-level annual average patient risk scores (Nosos scores) will be included to account for differences in relative comorbidity burden of facilities' patient populations, obtained from VSSC.¹⁶ For the specialty care wait times models only, we will include facility-level average enrollee driving distance, from PSSG, and community care wait times (for care outside of VA facilities), from CDW. We have conducted power analyses to determine the minimum detectable effect size for each outcome that can be detected with 80% power (Appendix G), which will be useful for putting our final results into context.

There may be baseline differences between the treatment and comparison groups on our control variables, described above and summarized in Table 4. If we observe a greater than 10% standardized difference, a commonly used threshold, we will also explore using coarsened exact matching on facility characteristics to pair treatment and comparison sites to improve our ability to provide a causal interpretation for our findings.¹⁷⁻²⁰ There is also the risk of randomization failure due to incomplete or less than ideal implementation of the intervention, including clinic drop out or difficulty in hiring and retention of scribes, in which case we will explore using randomization as an instrumental variable.

Implementation Results

Our results at this early stage focus on the implementation of the trial, which involves the hiring of 48 medical scribes in the 12 participating VAMCs as either VA employees or contractors. A key implementation outcome is the difference in hiring experiences between VA and contract scribes. Despite having reached the official start date, which itself was delayed due to COVID-19, hiring is still in progress and is expected to continue until all positions are filled, and thereafter as needed due to attrition. As the progression of the pandemic and hiring for the

pilot are unknown, we felt it worthwhile to report our findings despite the challenges we have encountered in implementing the trial. Our initial data (Figure 1) shows substantial differences in the hiring experience of VA versus contract scribes, with contracting outpacing VA hiring to date though with a nearly 50% longer time to hire per scribe. This graph depicts attrition after a scribe has reached their EOD date, not accounting for potential attrition for projected contract and VA scribes in the onboarding process. At this time, we do not anticipate that attrition during the onboarding process will significantly change these hiring projections.

Figure 1. Hiring of contract and employee scribes

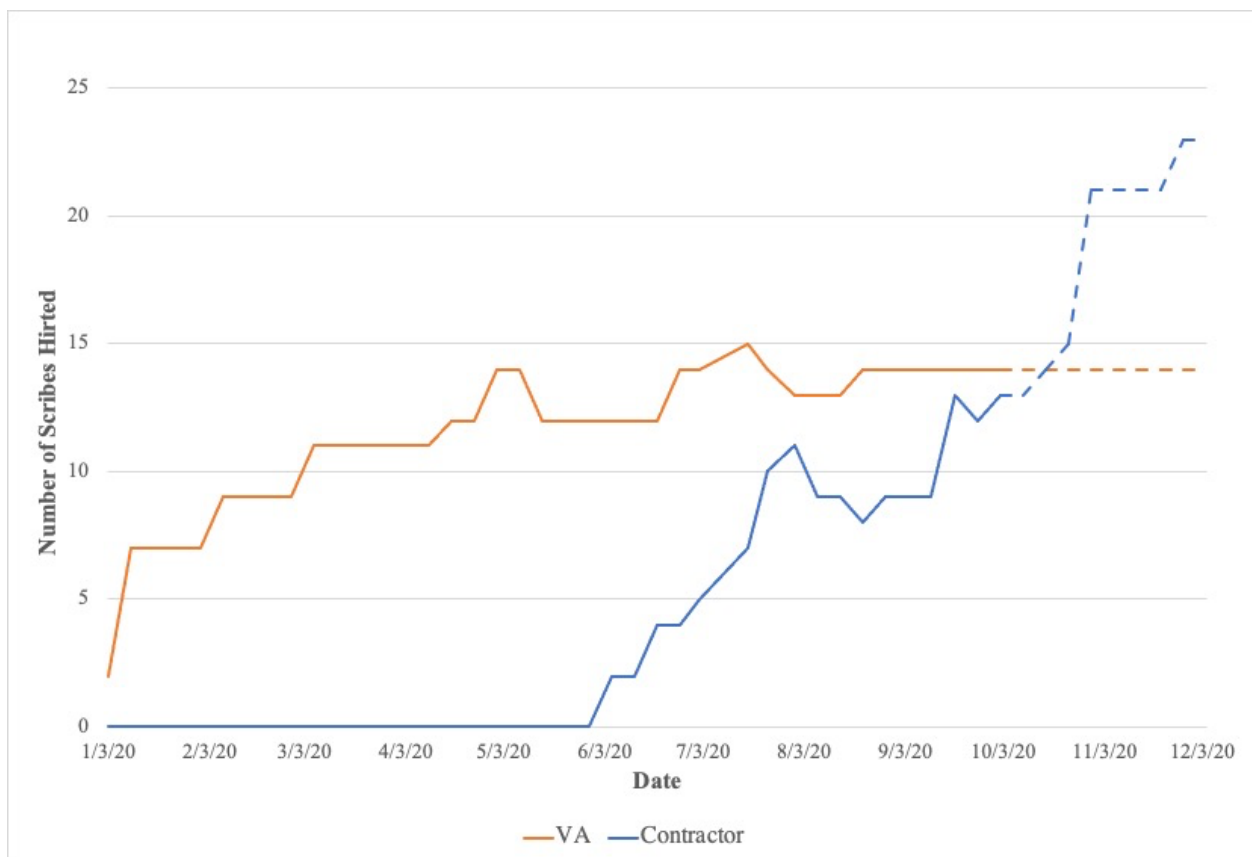


Figure 1 does not explicitly break down any difference in the time required to onboard the scribes through these two channels, which is an important consideration. For the scribes that have an EOD date, we also analyzed the differences in time to hire between employee and

contract scribes. We measured the time to hire as the time between the job being posted and EOD date, which includes the onboarding time for candidates. Employee scribes had an average time to hire of approximately 120 days versus approximately 198 days for contract scribes. Based on our initial observations of hiring to date, contracting seems to hold a significant advantage in terms of reaching its target but does seem to take somewhat longer when compared against the fewer completed employee hires. More data and a deeper analysis of the underlying causes of this variation in hiring trajectories between the employee and contract scribes will provide valuable insights if such scribes were to be deployed at scale within VHA.

Discussion

As a learning health care system and the largest integrated delivery system in the United States, VHA considers this trial to have great potential to increase our understanding of the potential effect of scribes while also informing a very real policy problem involving high wait times and high provider administrative burdens. Our evaluation findings from this pilot will provide insight into whether scribes can help increase provider productivity and decrease wait times for high demand specialties in the VA without adversely affecting patient satisfaction.

Implementation was complicated by the COVID-19 pandemic but a key takeaway thus far is that scribe hiring through contracting has been able to hit its target while VA hiring has fallen far short. Ongoing hiring results as the pilot continues, in filling the remaining positions and any attrition during the two-year study period, will help inform what effect COVID-19 had and if it differed between VA and contract hiring. Every 180 days, the VA must submit a report to Congress on the progress and impact of the program on provider efficiency, patient satisfaction, and average wait times. At the end of the study, the Comptroller General will submit a report comparing this program to similar programs conducted in the private sector. VACE will also be conducting a

separate qualitative analysis, not detailed here, to understand contextual factors impacting study outcomes. These reports will inform policy makers on the strengths and limitations of using VA and contract medical scribes as part of VHA care.

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