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Variation in Outpatient Antibiotic Prescribing for Acute Respiratory Infections in the Veteran Population: A Cross-sectional Study.

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1TITLE PAGE:

2“Variation in Outpatient Antibiotic Prescribing for Acute Respiratory Infections in the Veteran  
3Population.”

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34

35ABSTRACT

36

37BACKGROUND: Despite efforts to reduce antibiotic prescribing for acute respiratory infections  
38(ARIs), information on factors driving prescribing is limited.

39

40OBJECTIVE: To examine trends in antibiotic prescribing within the national Veterans Affairs  
41population over an 8-year period and to identify patient, provider, and setting sources of  
42variation.

43

44DESIGN: Retrospective cross-sectional study.

45

46SETTING: All emergency departments, primary care and urgent care clinics within the Veterans  
47Affairs health system.

48

49PARTICIPANTS: All visits between 2005 and 2012 with a primary diagnosis of acute respiratory  
50tract infection that typically have low proportions of bacterial infection. Patients with infections  
51or comorbidities indicating antibiotic use were excluded.

52

53MEASUREMENTS: Overall antibiotic prescription, macrolide prescription, patient, provider,  
54and setting characteristics extracted from the electronic medical record.

55

56RESULTS: Of 1.0 million ARI patient visits, the proportion resulting in an antibiotic prescription  
57increased from 67.5% in 2005 to 69.2% in 2012 ( $P<0.001$ ). The proportion of antibiotic  
58prescriptions that were macrolides increased from 36.8% to 47.0% ( $P<0.001$ ). Antibiotic  
59prescribing was highest for the diagnoses of sinusitis (adjusted proportion 86%) and bronchitis  
60(85%), while prescribing varied little by fever, age, setting type, or comorbidities. The majority  
61of variation in antibiotic use was attributable to providers: the highest 10% of providers  
62prescribed antibiotics for  $\geq 95\%$  of their patients, while the lowest 10% prescribed antibiotics to  
63 $\leq 40\%$ .

64

65LIMITATIONS: Retrospective. Lacked some clinical data that may influence the antibiotic  
66decision.

67

68CONCLUSIONS: Patients with ARIs commonly receive antibiotics in the VA population,  
69regardless of patient, provider, and setting features. The use of macrolides has increased. The  
70majority of variation in antibiotic use was at the provider level.

71

## 72INTRODUCTION

73 As the emergence of resistant pathogens outpaces our ability to develop new antibiotics,  
74the problem of unnecessary antibiotic use – a primary cause of the development of resistant  
75organisms – has become a major public health concern.(1,2) Despite limited benefits, the  
76majority of outpatient antibiotics are prescribed for acute respiratory infections (ARIs),(3,4) a  
77practice that is discouraged by practice guidelines.(5,6)

78 In response to this problem, local and national efforts across the United States have been  
79launched to improve prescribing behavior and develop criteria for using antibiotics to treat  
80respiratory infections. While there has been a significant decline in overall antibiotic use for  
81ARIs for children,(7) the use of antibiotics in adults remains high, and recent studies demonstrate  
82a dramatic increase in the use of broad-spectrum antibiotics, particularly macrolides.(8,9,10)  
83Tracking national practice patterns and identifying sources of variation in antibiotic use would  
84improve our ability to target interventions more appropriately. Previous studies have identified  
85facility-level variation in prescribing patterns(11) and differences based upon patient and  
86provider characteristics;(12,13) however, these associations are difficult to interpret without  
87analytic techniques that take into consideration the effects of different levels in healthcare  
88delivery (provider, clinic and healthcare system).

89 The aims of this study were to: 1) measure national trends in antibiotic prescriptions for  
90ARIs at outpatient facilities within the Veterans Affairs Health System during an eight-year  
91period; 2) investigate patient, provider and setting factors associated with the prescription of any  
92antibiotic as well as a macrolide; and 3) measure variation at the provider, clinic, and medical  
93center levels.

94

## 95METHODS:

### 96Setting

97 The Veterans Affairs network serves nearly 8.5 million Veterans each year at over 1700  
98clinics and 152 hospitals, with approximately 13 million primary care visits per year.(14) Where  
99multiple hospital divisions operate as an integrated healthcare system under a single leadership  
100team, these facilities are combined, resulting in 139 distinct VA Medical Centers (VAMCs). The  
101primary care needs of Veterans are met in primary care, urgent care and emergency department  
102(ED) settings across the VA system. These settings are either located on the physical grounds of a  
103VA Medical Center and its local hospital (VAMC-based), or located at Community-Based  
104Outpatient Clinics (CBOCs), which are stand-alone facilities that offer outpatient services only.

105 All healthcare settings within the VA share the same clinical electronic health record. All  
106data for our study were accessed through the Veterans Health Information Systems and  
107Technology Architecture (VistA), through the Veterans Informatics and Computing Infrastructure  
108(VINCI), a computing environment that stores clinical data for research purposes.(15,16)

### 109Participants

110 During the period of January 1, 2005, to December 31, 2012, we identified all patient  
111visits at emergency departments, primary care, and urgent care clinics with *International*  
112*Classification of Disease-Edition 9* (ICD-9) codes consistent with acute respiratory tract  
113infections, including nasopharyngitis (460), pharyngitis (462), sinusitis (461.x), acute bronchitis  
114(466.x), upper respiratory tract infection (465.8, 465.9), and other infections such as laryngitis  
115(464) and tonsillitis (463). Visits with ICD-9 codes for skin or soft tissue infection, pneumonia,  
116influenza, urinary tract infection, or other infections at the same visit were excluded, as were  
117patients with a previous acute respiratory infection within the past 30 days. We also excluded

118patients with ICD-9 codes consistent with comorbidities that increase the risk of serious bacterial  
119infections, including HIV, neoplasms, diabetes, chronic lung disease including COPD and  
120asthma, end stage renal failure requiring hemodialysis, solid organ transplantations, or other  
121immunocompromised states within one year of the visit. A complete list of ICD-9 codes used for  
122inclusion and exclusion is listed in Appendix A.

123

124*Measurements: patient, provider, and setting factors*

125 We extracted all patient, provider, and setting features listed in Table 1. Although patients  
126with comorbidities that increase risk of serious bacterial infections were excluded, we estimated  
127the burden of other comorbidities for each included patient by extracting all ICD-9 codes given  
128to each patient within the year prior to the ARI visit, and applied the Clinical Classification  
129System (CCS) developed by the Agency for Health care Research and Quality.<sup>(17)</sup> We further  
130grouped the CCS into six categories relevant to the ARI diagnosis (infection, pulmonary, renal,  
131cardiovascular, psychiatric, and immunodeficient disease). Patient distance to the facility was  
132measured by calculating the travel distance between the patient’s home address and the location  
133of the visit.

134 We identified a single “primary provider” reported for the day of each visit. Although  
135there could be multiple providers for multiple visits in one day, the primary provider is the single  
136health professional who identifies him/herself in the electronic health record as the individual  
137who is responsible for the decision-making, patient care, and documentation at that particular  
138encounter. The primary provider for the encounter was not necessarily the *primary care provider*  
139(PCP) for the patient’s general medical care; for example, if the patient presented to the  
140emergency department, the “primary provider” for the visit would likely be a provider other than

141the PCP. Additionally, the primary provider listed was not necessarily the prescriber of  
142medications, as the electronic order entry of medications was not necessarily performed by the  
143same individual who completed the documentation of the visit. However, this was the case for  
14490% of all listed physicians and mid-level providers.

145 We did not classify providers by level of training (residents or other trainees), as the  
146attending physician rather than trainee was usually listed as the primary provider. Due to  
147incomplete provider census information, we did not measure specialty or years since licensure.  
148Patient visits that lacked documentation of a temperature (16%) were categorized as having no  
149fever. Missing provider age (22%) was included in the model as a separate category. The  
150remainder of the visits that contained missing values (totaling less than 1%) were excluded from  
151the analysis.

152

### 153*Measurements: antibiotic prescriptions*

154 We initially extracted all VA antibiotic prescription fills within 2 days prior and 3 days  
155after the patient visit. As it was possible for patients to receive antibiotic prescriptions over the  
156phone prior to a visit or to fill a prescription after the visit occurred, we chose this date range to  
157identify all prescription fills. Because some VA facilities lacked fill data due to outside pharmacy  
158services as well as variation in data collection, we then applied RED, a natural language  
159processing tool developed by VINCI,(18) to identify antibiotic prescriptions within unstructured  
160clinical documents. The resulting RED algorithm demonstrated a positive predictive value of  
16198% against a reference standard of physician review. After text classification, the proportion of  
162patient visits that were identified to result in an antibiotic prescription increased from 60.7% to  
16368.4%.

164

165 *Statistical Analysis*

166 Annual trends in overall antibiotic prescriptions as well as macrolide prescriptions were  
167 tested for significance using univariate logistic regression with the calendar year as the single  
168 linear predictor variable.

169 *Relationship between antibiotic prescription and patient, provider and setting characteristics*

170 We used generalized estimating equations (GEE) under multivariable linear logistic  
171 regression models to predict the probability of 1) prescribing any antibiotic at a patient visit, and  
172 2) prescribing a macrolide when an antibiotic was prescribed. The following predictors were  
173 used: patient gender, patient age, number of comorbidities, maximum temperature, diagnosis,  
174 distance to clinic, type of listed provider, provider age, daily ARI visit load, time of day, calendar  
175 month, and calendar year. Calendar year was modeled as a linear predictor of antibiotic  
176 prescribing. All other continuous variables were categorized into quartiles as cutoff points except  
177 for temperature, which had clinically defined cutoffs (“high fever” = temperature  $\geq 102F$ ;  
178 “fever” = temperature  $100.4F$  and  $< 102F$ ), daily clinic load (1, 2, 3, and  $\geq 4$  ARI visits per day),  
179 and time of day (early morning 0800-10:30, late morning 10:30-12:30, early afternoon 12:30-  
180 14:30, late afternoon 14:30-17:00, and night 17:00-08:00). GEE was used for this portion of the  
181 analysis to generate population weighted average comparisons and to provide statistical  
182 inferences that were more robust to potential misspecification of the model used to account for  
183 clustering by VAMC, clinic and provider. The GEE analyses were performed with an  
184 independent working covariance model for encounters within the same VAMC to assure that  
185 each encounter was weighted equally in each analysis. We calculated the mean adjusted  
186 proportions using the marginal standardization approach, in which adjusted proportions are



187summed to a weighted average reflecting the distribution of the remaining predictor variables in  
188the target population.(19) Bootstrapping with 400 independent samples drawn from the 130  
189VAMCs was utilized to provide 95% confidence intervals and p-values for the adjusted  
190proportions.

191

### 192*Variation in antibiotic prescribing associated with different levels of healthcare delivery*

193 We fit a generalized linear mixed effect models for antibiotic prescribing, with provider,  
194clinic/ED, and VAMC included as normally distributed nested random effects on the logit scale,  
195and with fixed effects terms representing the above listed patient, provider, and setting factors.  
196The estimated variances for provider, clinic/ED, and VAMC were used to describe the variation  
197in antibiotic prescribing specifically attributable to each of these levels of the healthcare system,  
198after controlling for the fixed effects.

199 Because providers are not perfectly nested within clinic, providers who appeared in  
200multiple clinics were reassigned to the unique clinics in which they appeared most frequently. In  
201order to simplify computations and compare the modeled results with the observed variation in  
202the prevalence of antibiotic prescription across providers, we restricted this analysis to those  
203providers with at least 100 patients in the study, representing a total of 480,875 visits.

204 We used three approaches to visualize the different levels of variation in antibiotic  
205prescribing. We first generated a histogram displaying the observed, unadjusted distribution of  
206antibiotic prescribing proportions across providers. Second, we used the results of the  
207generalized mixed model to estimate density functions describing the total variation in  
208prescribing attributable to each of the three random effects. We also displayed the density  
209function with variance given by the sum of the estimated variance of the three random effects to

210describing the total variation in prescribing, incorporating variation from all three levels of  
211healthcare delivery.

212 Third, we displayed conditional density curves to display the conditional distributions of  
213antibiotic prescription prevalence across providers within clinics with prescribing prevalence  
214fixed at the 10<sup>th</sup>, 50<sup>th</sup>, and 90<sup>th</sup> percentiles of the distribution of antibiotic prescribing across  
215clinics, and within VAMCs with prescription prevalence fixed at the 10<sup>th</sup>, 50<sup>th</sup>, and 90<sup>th</sup>  
216percentiles of the distribution of antibiotic prescribing across VAMCs.(20)

217 The study was conducted with approval from the University of Utah Institutional Review  
218Board (IRB#00058373) and the VA, and with support from the Centers for Disease Control and  
219Prevention, who participated in study design and interpretation of the results. Statistical analyses  
220were performed using STATA 12.0 and SAS 9.2. All analysis code is documented in Appendix C.

221

## 222RESULTS

### 223*Study Population*

224 Of 2,481,520 total patient visits with a diagnosis of ARI, we identified 1,044,523 that met  
225our inclusion criteria (Figure 1). These were staffed by 45,619 providers at 990 clinics or  
226emergency departments within 130 VAMCs. Thirty percent of all visits occurred in community-  
227based outpatient clinics, while the remainder occurred in clinics/EDs on VAMC campuses.  
228Seventy percent occurred in primary care clinics, 23% in emergency departments, and 7% in  
229urgent care clinics. A physician was listed as the primary provider for 62% of visits, followed by  
230mid-level provider (24%) and nurse (11%). The median age of the patient population was 61;  
23151.9% had documentation of a cardiovascular comorbidity, and 24.1% had documentation of a

232pulmonary comorbidity not included in the exclusion criteria. Twenty-five percent of the  
233population lived >31 miles from their visit location.

234

#### 235*National Trends*

236 We observed a small, unadjusted absolute increase in overall proportion of visits with  
237antibiotic prescriptions during the study period from 67.5% in 2005 to 69.2% in 2012 ( $P<0.001$ ).  
238Although we observed a seasonal trend in the number of ARI-related visits, there was no  
239substantial seasonal variation in the proportion of these visits for which antibiotics were  
240prescribed (Figure 2). Of the visits resulting in an antibiotic prescription, macrolide prescriptions  
241increased from 36.8% in 2005 to 47.0% in 2012 ( $P<0.001$ ), while penicillins (36% to 32.1%,  
242 $P<0.001$ ) and fluoroquinolones (15.0% to 12.7%,  $P<0.001$ ) decreased over time (Figure 3).

243

#### 244*Predictors of antibiotic prescribing*

245 Antibiotics were prescribed in 68.4% of all ARI visits ( $N=714,552$ ). Figure 4 displays  
246adjusted proportions of visits with antibiotic prescription for subgroups defined by selected  
247factors. Subgroups associated with higher prevalence of antibiotic prescribing included a  
248diagnosis of sinusitis (adjusted percent 87%) or bronchitis (85%), presence of a high fever  
249(78%), occurrence in an urgent care setting (77%), southern region (72%), and central region  
250(72%). Mid-level providers also prescribed antibiotics slightly more frequently than physicians  
251(71% versus 69%). “Other” providers also had a higher prescribing prevalence (80%), likely due  
252to the high proportion of pharmacists in this small group (<2% of all visits). Additionally,  
253prescribing was slightly higher at clinics based in VA medical centers than at community-based

254outpatient clinics (71% versus 63%). The number of patient comorbidities that were not in the  
255exclusion criteria had no association with antibiotic prescribing.

256 Of the antibiotics prescribed, 43.6% (N=302,595) were macrolides. Subgroups with  
257elevated adjusted prevalence estimates for macrolide prescribing (Figure 5) included a diagnosis  
258of bronchitis (adjusted percent 52%) or URI (50%). Presence of a high fever was a negative  
259predictor for macrolide use (36%). No clinically significant differences were seen in other patient  
260features, provider features, or geographic region.

261

262*Sources of variation.*

263 The histogram in Figure 6a displays the variation in the proportions of visits with  
264antibiotic prescriptions among providers who saw more than 100 patients with ARIs during the  
265study period (N=480,875 visits, seen by 2,594 providers). The highest 10% providers prescribed  
266antibiotics for at least 95% of their patients, while the lowest 10% prescribed antibiotics to  $\leq$   
26740% of their patients. The smooth curve in the same figure displays the modeled variation in  
268prescribing among providers after adjustment for the previously described measured patient,  
269provider and setting characteristics. The similarity of the curve to the histogram suggests that  
270these measured characteristics contributed only slightly to the overall variation in antibiotic  
271prescribing across providers.

272 The dashed curves in Figure 6b display the model-generated estimates of variation  
273specifically associated with each level of healthcare delivery (provider, clinic, and VAMC), while  
274controlling for the measured patient, provider, and setting fixed effect factors. After accounting  
275for the fixed effect factors, variation attributable to providers, clinics, and VAMCs respectively  
276accounted for 59%, 28% and 13% of the total remaining variation in antibiotic prescription

277prevalence across the three levels of healthcare delivery. For comparison, the solid curve in the  
278same figure (which matches the solid curve in Figure 6a) displays the model-based overall  
279variation in prescribing among providers. The figure of Appendix B displays how the modeled  
280distributions of prescribing across providers varies between clinics and VAMCs. Even within  
281high or low prescribing VAMCs and clinics, there was considerable variation in provider  
282prescribing.

283

#### 284DISCUSSION:

285       Our 8-year study of 1 million patient visits demonstrates a persistently high prevalence of  
286outpatient antibiotic prescriptions for acute respiratory infections in the national Veteran  
287population. During the same period, macrolides have become the predominant class prescribed.  
288Similar trends have been reported in studies using data from the National Ambulatory Medical  
289Care Survey (NAMCS) and National Hospital Ambulatory Medical Care Survey (NHAMCS).  
290(21, 22, 23, 24) The lack of progress in reducing unnecessary antibiotic prescribing for ARIs is a  
291major public health concern.

292       We aimed to gain a better understanding of the factors driving this problem. Our  
293exceptionally large population of patients, providers, and clinics within the VA health system  
294gave us the unique opportunity to characterize variation in antibiotic prescribing across different  
295levels of healthcare delivery. The granularity of our data further enabled us to explore  
296relationships between antibiotic use and multiple factors, including patient characteristics,  
297provider experience, and clinic features.

298       Antibiotic prescribing was associated with many of the factors we measured, including  
299temperature, distance to clinic, setting type, and geographic region. These associations, however,

300were small, and even when taken together, they had limited explanatory power. Antibiotic  
301prescribing was common regardless of the patient and setting features we studied.

302       The greatest source of variability in management we identified was the provider. While  
30310% of our providers prescribed antibiotics for at least 95% of all of their ARI visits, another  
30410% prescribed antibiotics for 40% or less. After adjustment of all of the factors studied, we  
305discovered a magnitude of variation at the provider level that overshadowed the clinic or medical  
306center.

307       Other studies of antibiotic use in ARIs have identified variation at the facility,(11) health  
308plan,(25) and regional(26) levels, and cultural influences and context in both provider practices  
309and patient expectations have been provided as important reasons for environmental influences  
310on antibiotic use.(27) Indeed, much medical decision-making is influenced by the system and  
311social context. While environment is important, our findings suggest that providers have a strong  
312tendency to choose the same treatment decision regardless of patient or clinic features, indicating  
313that individual provider preference, or “style,” exerts a heavy influence on the antibiotic  
314decision.

315       We found a substantial increase in macrolide prescribing in our system, a trend that has  
316been observed in other national studies.(9),(10) The increase in popularity of macrolides could  
317be due to its short course, convenient, once-daily dosing, the recommendation of macrolides as  
318first-line empiric treatment for community-acquired pneumonia, and successful marketing  
319campaigns. Macrolides are not recommended as first line therapy for either pharyngitis or  
320sinusitis, and while we observed lower proportions of macrolide use for these diagnoses than  
321bronchitis and URIs, the proportion was still significant. This trend is concerning given the lack  
322of additional benefit of macrolides over narrow-spectrum antibiotics for the treatment of ARI’s,

323the increase in macrolide-resistant pneumococcal disease,(28) and potential cardiotoxicity,(29)  
324especially considering the high proportion of Veterans in our study that had cardiovascular  
325comorbidities.

326 We recognize several limitations to our study. We used administrative data, relying upon  
327ICD-9 codes to identify our population. Additionally, we excluded a large number of patients  
328with comorbidities and diagnoses for infections with greater risk for bacterial infections in an  
329attempt to identify cases with a low risk for bacterial infection (and hence a lower likelihood to  
330benefit from antibiotics). For any individual patient, whether an antibiotic was appropriate is  
331impossible to determine. However, at the population level, we would expect a much lower  
332overall proportion of antibiotic prescribing for this group of patients based upon national  
333treatment guidelines.

334 Due to the diversity of settings studied and variation in charting practices, our study  
335lacked additional clinical data that have been previously shown to play a role the antibiotic  
336decision, such as symptom duration or physical exam findings, or provider information such as  
337specialty or training background.(30, 31) While we attempted to exclude patients who had  
338comorbidities that would increase risk for serious bacterial infections, different providers have  
339different patient panels, and thus different thresholds for antibiotic prescribing. However, the  
340provider-level variation that we observed remained after adjusting for additional patient  
341comorbidities and other clinical features, and providers varied widely within clinics, suggesting  
342against a significant amount of tailoring based on measurable patient factors. Additional factors  
343that might explain the degree of provider variation in antibiotic prescriptions, such as attitudes  
344toward the risks and benefits of antibiotics, responsiveness to local surveillance data or  
345stewardship efforts, or patient preferences and expectations (and providers' understanding of

346them), could potentially be identified on a large scale within the clinical record in the future.

347Further research that incorporates qualitative methods will also improve our ability to elucidate  
348these mechanisms.

349 Our Veteran population includes a greater proportion of older, male patients with a higher  
350comorbidity burden. Despite this, we found similar overall proportions, geographic differences,  
351and trends in macrolide use to those in other studies. Ambulatory care and pharmacy services at  
352the VA are more integrated and standardized than other ambulatory settings. In other more  
353diverse practice settings, clinic- or system-based variations in infrastructure might play a larger  
354role in driving differences in antibiotic use. However, the provider-level variation that we  
355observed might also be even greater in other systems.

356 Our findings have important implications for health systems and public health. Variation  
357in management of acute respiratory infection does not appear to be driven by tailoring of  
358treatment to an individual patient's circumstances but rather by prescribing patterns of individual  
359providers. As a prime example of unexplained variation, this is a ripe target for quality  
360improvement and antibiotic stewardship interventions. Audit and feedback has shown promise as  
361a powerful tool to change behavior,(32, 33) and new provider-targeted decision support tools  
362could help clinicians recognize and respond to their prescribing patterns compared to similar  
363provider and patient populations.

364 Unnecessary antibiotic use for acute respiratory infections remains an important problem.  
365The persistence of this problem requires new approaches. As our understanding of the  
366relationship between providers, patients, settings and treatment decisions improves, so will our  
367ability to target future information and stewardship efforts.

368



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375Prevention, or the United States government.

376

377REPRINT REQUESTS

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384

385 **TABLE AND FIGURES**

386 Table 1. Patient, provider and setting characteristics in ARI visits for the Veteran population.  
387

388 Figure 1. Study population.

389 Visits could meet multiple exclusion criteria, and thus the sum of visits meeting each exclusion  
390 criteria exceeds the total number of excluded visits.  
391

392 Figure 2. Trends in overall antibiotic prescribing.

393 Number of ARI-related visits per month and monthly proportion of visits resulting in an  
394 antibiotic prescription.  
395

396 Figure 3. Temporal trends in the proportion of all antibiotics prescribed for each antibiotic class.  
397

398 Figure 4. Predictors of antibiotic prescribing.

399 For each subgroup, the adjusted proportion of visits with antibiotics prescribed based on the  
400 marginal standardization model is shown. N = 1,036,982 visits. Model also included calendar  
401 month and year. Statistically significant predictors ( $p < 0.001$ ) included the diagnosis of sinusitis  
402 or bronchitis, fever or high fever, provider type of "other", urgent care clinic, daily clinic ARI  
403 load > 4 visits, central region, and south region.  
404

405 Figure 5. Predictors of macrolide prescribing.

406 For each subgroup, adjusted proportion of antibiotic prescriptions that were macrolides is shown.  
407 N = 714,552 visits. Model also included calendar month and year. Statistically significant  
408 predictors ( $p < 0.001$ ) included the diagnosis of URI, bronchitis, and normal temperature.  
409

410 Figure 6. Variation in antibiotic prescribing.

411 Figure 6a. Variation in antibiotic prescribing among providers.

412 Histogram displays the distribution of observed proportions of visits with an antibiotic  
413 prescription across 2,594 providers with at least 100 ARI visits each (total N=480,875). Smooth  
414 curve depicts the modeled distribution of antibiotic prescription across providers, after  
415 controlling for measured patient, provider, and setting features listed in Figures 4 and 5.  
416

417 Figure 6b. Sources of variation in antibiotic prescribing.

418 Dashed curves depict modeled distributions describing variation in antibiotic prescription  
419 proportions attributable specifically to providers, clinics and VAMCs after controlling for the  
420 measured patient, provider and setting parameters listed in Figures 4 and 5. Solid curve  
421 corresponds to the solid curve in Figure 6a, and depicts overall modeled variation in antibiotic  
422 prescription across providers, including differences between providers at different clinics and  
423 VAMCs.  
424

425 APPENDIX A. Definitions of inclusion, exclusion criteria and diagnostic code groupings.

426

427 APPENDIX B. Figure. Conditional distribution of antibiotic prescribing.

428 Conditional density curves of antibiotic prescribing prevalence across providers within clinics  
429 with prescription prevalence fixed at the 10<sup>th</sup>, 50<sup>th</sup>, and 90<sup>th</sup> percentiles of the distribution of

430antibiotic prescribing across clinics, within VAMCs with prescribing prevalence fixed at the 10<sup>th</sup>,  
43150<sup>th</sup>, and 90<sup>th</sup> percentiles of the distribution of antibiotic prescribing, across VAMCs.

432

433APPENDIX C. Analysis code, statistical appendix, and separation of between-cluster and within-  
434cluster variables.

435

436

437

438

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