AHRQ Quality Indicators

Guide to Prevention Quality Indicators:
Hospital Admission for
Ambulatory Care Sensitive Conditions

Department of Health and Human Services
Agency for Healthcare Research and Quality
http://www.qualityindicators.ahrq.gov

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Preface

In health care as in other arenas, that which cannot be measured is difficult to improve. Providers, consumers, policy makers, and others seeking to improve the quality of health care need accessible, reliable indicators of quality that they can use to flag potential problems, follow trends over time, and identify disparities across regions, communities, and providers. As noted in a 2001 Institute of Medicine study, *Envisioning the National Health Care Quality Report*, it is important that such measures cover not just acute care but multiple dimensions of care: staying healthy, getting better, living with illness or disability, and coping with the end of life.

The Agency for Healthcare Research and Quality (AHRQ) Quality Indicators (QIs) are one Agency response to this need for a multidimensional, accessible family of quality indicators. They include a family of measures that providers, policy makers, and researchers can use with inpatient data to identify apparent variations in the quality of either inpatient or outpatient care. AHRQ’s Evidence-Based Practice Center (EPC) at the University of California San Francisco (UCSF) and Stanford University adapted, expanded, and refined these indicators based on the original Healthcare Cost and Utilization Project (HCUP) Quality Indicators developed in the early 1990s.

The new AHRQ QIs are organized into four modules: Prevention Quality Indicators (PQIs), Inpatient Quality Indicators (IQIs), Patient Safety Indicators (PSIs), and Pediatric Quality Indicators (PDIs). AHRQ has published the modules as a series. Full technical information on the first two modules can be found in *Refinement of the HCUP Quality Indicators (Summary), May 2001* prepared by the UCSF-Stanford EPC. It can be accessed at [http://www.qualityindicators.ahrq.gov/downloads.htm](http://www.qualityindicators.ahrq.gov/downloads.htm).

This first module focuses on preventive care services—outpatient services geared to staying healthy and living with illness. Researchers and policy makers have agreed for some time that inpatient data offer a useful window on the quality of preventive care in the community. Inpatient data provide information on admissions for “ambulatory care sensitive conditions” that evidence suggests could have been avoided, at least in part, through better outpatient care. Hospitals, community leaders, and policy makers can then use such data to identify community need levels, target resources, and track the impact of programmatic and policy interventions.

One of the most important ways we can improve the quality of health care in America is to reduce the need for some of that care by providing appropriate, high-quality preventive services. For this to happen, however, we need to be able to track not only the level of outpatient services but also the outcome of the services people do or do not receive. The PQIs are intended to facilitate such efforts. The PQIs are already being applied at the national level in the National Healthcare Quality Report ([http://qualitytools.ahrq.gov/qualityreport](http://qualitytools.ahrq.gov/qualityreport)) and National Healthcare Disparities Report ([http://qualitytools.ahrq.gov/disparitiesreport](http://qualitytools.ahrq.gov/disparitiesreport)). As always, we would appreciate hearing from those who use our measures and tools so that we can identify how they are used, how they can be refined, and how we can measure and improve the quality of the tools themselves.

Irene Fraser, Ph.D., Director
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The programs for the Prevention Quality Indicators (PQIs) can be downloaded from [http://www.qualityindicators.ahrq.gov/pqi_download.htm](http://www.qualityindicators.ahrq.gov/pqi_download.htm). Instructions on how to use the programs to calculate the PQI rates are contained in the companion text, *Prevention Quality Indicators: Software Documentation (SAS or SPSS) or AHRQ QI Windows Application Documentation.*
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1.0 Introduction to the AHRQ Prevention Quality Indicators

Prevention is an important role for all health care providers. Providers can help individuals stay healthy by preventing disease, and they can prevent complications of existing disease by helping patients live with their illnesses. To fulfill this role, however, providers need data on the impact of their services and the opportunity to compare these data over time or across communities. Local, State, and Federal policymakers also need these tools and data to identify potential access or quality-of-care problems related to prevention, to plan specific interventions, and to evaluate how well these interventions meet the goals of preventing illness and disability.

The Agency for Healthcare Research and Quality (AHRQ) Prevention Quality Indicators (PQIs) represent one such tool. Local, State, or national data collected using the PQIs can flag potential problems resulting from a breakdown of health care services by tracking hospitalizations for conditions that should be treatable on an outpatient basis, or that could be less severe if treated early and appropriately. The PQIs represent the current state of the art in measuring the outcomes of preventive and outpatient care through analysis of inpatient discharge data.

1.1 What Are the Prevention Quality Indicators?

The PQIs are a set of measures that can be used with hospital inpatient discharge data to identify "ambulatory care sensitive conditions" (ACSCs). ACSCs are conditions for which good outpatient care can potentially prevent the need for hospitalization, or for which early intervention can prevent complications or more severe disease.

Even though these indicators are based on hospital inpatient data, they provide insight into the quality of the health care system outside the hospital setting. Patients with diabetes may be hospitalized for diabetic complications if their conditions are not adequately monitored or if they do not receive the patient education needed for appropriate self-management. Patients may be hospitalized for asthma if primary care providers fail to adhere to practice guidelines or to prescribe appropriate treatments. Patients with appendicitis who do not have ready access to surgical evaluation may experience delays in receiving needed care, which can result in a life-threatening condition—perforated appendix. The PQIs consist of the following 14 ambulatory care sensitive conditions, which are measured as rates of admission to the hospital:

<table>
<thead>
<tr>
<th>PQI Number</th>
<th>Prevention Quality Indicators</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Diabetes short-term complication admission rate</td>
</tr>
<tr>
<td>2</td>
<td>Perforated appendix admission rate</td>
</tr>
<tr>
<td>3</td>
<td>Diabetes long-term complication admission rate</td>
</tr>
<tr>
<td>4</td>
<td>Chronic obstructive pulmonary disease admission rate</td>
</tr>
<tr>
<td>5</td>
<td>Hypertension admission rate</td>
</tr>
<tr>
<td>6</td>
<td>Congestive heart failure admission rate</td>
</tr>
<tr>
<td>7</td>
<td>Low Birth Weight</td>
</tr>
<tr>
<td>8</td>
<td>Dehydration admission rate</td>
</tr>
<tr>
<td>9</td>
<td>Bacterial pneumonia admission rate</td>
</tr>
<tr>
<td>10</td>
<td>Urinary tract infection admission rate</td>
</tr>
<tr>
<td>11</td>
<td>Angina admission without procedure</td>
</tr>
<tr>
<td>12</td>
<td>Uncontrolled diabetes admission rate</td>
</tr>
<tr>
<td>13</td>
<td>Adult asthma admission rate</td>
</tr>
<tr>
<td>14</td>
<td>Rate of lower-extremity amputation among patients with diabetes</td>
</tr>
</tbody>
</table>
PQIs #4 and #6 have been moved to the Pediatric Quality Indicators module. All PQIs now apply only to adult populations.

Although other factors outside the direct control of the health care system, such as poor environmental conditions or lack of patient adherence to treatment recommendations, can result in hospitalization, the PQIs provide a good starting point for assessing quality of health services in the community. Because the PQIs are calculated using readily available hospital administrative data, they are an easy-to-use and inexpensive screening tool. They can be used to provide a window into the community—to identify unmet community health care needs, to monitor how well complications from a number of common conditions are being avoided in the outpatient setting, and to compare performance of local health care systems across communities.

1.2 How Can the PQIs Be Used in Quality Assessment?

While these indicators use hospital inpatient data, their focus is on outpatient health care. Except in the case of patients who are readmitted soon after discharge from a hospital, the quality of inpatient care is unlikely to be a significant determinant of admission rates for ambulatory care sensitive conditions. Rather, the PQIs assess the quality of the health care system as a whole, and especially the quality of ambulatory care, in preventing medical complications. As a result, these measures are likely to be of the greatest value when calculated at the population level and when used by public health groups, State data organizations, and other organizations concerned with the health of populations.

These indicators serve as a screening tool rather than as definitive measures of quality problems. They can provide initial information about potential problems in the community that may require further, more in-depth analysis. Policy makers and health care providers can use the PQIs to answer questions such as:

- Does the admission rate for diabetes complications in my community suggest a problem in the provision of appropriate outpatient care to this population?
- How does the admission rate for congestive heart failure vary over time and from one region of the country to another?

State policy makers and local community organizations can use the PQIs to assess and improve community health care. For example, an official at a State health department wants to gain a better understanding of the quality of care provided to people with diabetes in her State. She selects the four PQIs related to diabetes and applies the statistical programs downloaded from the AHRQ Web site to hospital discharge abstract data collected by her State.

Based on output from the programs, she examines the age- and sex-adjusted admission rates for these diabetes PQIs for her State as a whole and for communities within her State. The programs provide output that she uses to compare different population subgroups, defined by age, ethnicity, or gender. She finds that admission rates for short-term diabetes complications and uncontrolled diabetes are especially high in a major city in her State and that there are differences by race/ethnicity. She also applies the PQI programs to multiple years of her State’s data to track trends in hospital admissions over time. She discovers that the trends for these two PQIs are increasing in this city but are stable in the rest of the State. She then compares the figures from her State to national and regional averages on these PQIs using HCUPnet—an online query system providing access to statistics based on HCUP data. The State average is slightly higher than the regional and national averages, but the averages for this city are substantially higher.

1 Individual hospitals that are sole providers for communities and that are involved in outpatient care may be able to use the PQI programs. Managed care organizations and health care providers with responsibility for a specified enrolled population can use the PQI programs but must provide their own population denominator data.
2 HCUPnet can be found at http://hcup.ahrq.gov/HCUPnet.asp and provides instant access to national and regional data from the Healthcare Cost and Utilization Project, a Federal-State-industry partnership in health data maintained by the Agency for Healthcare Research and Quality.
After she has identified disparities in admission rates in this community and in specific patient groups, she further investigates the underlying reasons for those disparities. She attempts to obtain information on the prevalence of diabetes across the State to determine if prevalence is higher in this city than in other communities. Finding no differences, she consults with the State medical association to begin work with local providers to discern if quality-of-care problems underlie these disparities. She contacts hospitals and physicians in this community to determine if community outreach programs can be implemented to encourage patients with diabetes to seek care and to educate them on lifestyle modifications and diabetes self-management. She then helps to develop specific interventions to improve care for people with diabetes and reduce preventable complications and resulting hospitalizations.

### 1.3 What does this Guide Contain?

This guide provides background information on the PQIs. First, it describes the origin of the entire family of AHRQ Quality Indicators. Second, it provides an overview of the methods used to identify, select, and evaluate the AHRQ Quality Indicators. Third, the guide summarizes the PQIs specifically, describes strengths and limitations of the indicators, documents the evidence that links the PQIs to the quality of outpatient health care services, and then provides in-depth two-page descriptions of each PQI.

Appendices A and B have been removed. The new document *Prevention Quality Indicators Technical Specifications* outlines the specific definitions of each PQI, with complete ICD-9-CM coding specifications. A new section, “Using Different Types of QI Rates,” explains the various types of rates calculated by the software and presents tips on selecting the appropriate type of rate to use for given situations.

See Appendix A for links to these and other documents as well as Web sites that may be of interest to PQI users.
2.0 Origins and Background of the Quality Indicators

In the early 1990s, in response to requests for assistance from State-level data organizations and hospital associations with inpatient data collection systems, AHRQ developed a set of quality measures that required only the type of information found in routine hospital administrative data—diagnoses and procedures, along with information on patient’s age, gender, source of admission, and discharge status. These States were part of the Healthcare Cost and Utilization Project, an ongoing Federal-State-private sector collaboration to build uniform databases from administrative hospital-based data.

AHRQ developed these measures, called the HCUP Quality Indicators, to take advantage of a readily available data source—administrative data based on hospital claims—and quality measures that had been reported elsewhere. The 33 HCUP QIs included measures for avoidable adverse outcomes, such as in-hospital mortality and complications of procedures; use of specific inpatient procedures thought to be overused, underused, or misused; and ambulatory care sensitive conditions.

Although administrative data cannot provide definitive measures of health care quality, they can be used to provide indicators of health care quality that can serve as the starting point for further investigation. The HCUP QIs have been used to assess potential quality-of-care problems and to delineate approaches for dealing with those problems. Hospitals with high rates of poor outcomes on the HCUP QIs have reviewed medical records to verify the presence of those outcomes and to investigate potential quality-of-care problems. For example, one hospital that detected high rates of admissions for diabetes complications investigated the underlying reasons for the rates and established a center of excellence to strengthen outpatient services for patients with diabetes.

2.1 Development of the AHRQ Quality Indicators

Since the original development of the HCUP QIs, the knowledge base on quality indicators has increased significantly. Risk-adjustment methods have become more readily available, new measures have been developed, and analytic capacity at the State level has expanded considerably. Based on input from current users and advances to the scientific base for specific indicators, AHRQ funded a project to refine and further develop the original QIs. The project was conducted by the UCSF-Stanford EPC.

The major constraint placed on the UCSF-Stanford EPC was that the measures could require only the type of information found in hospital discharge abstract data. Further, the data elements required by the measures had to be available from most inpatient administrative data systems. Some State data systems contain innovative data elements, often based on additional information from the medical record. Despite the value of these record-based data elements, the intent of this project was to create measures that were based on a common denominator discharge data set, without the need for additional data collection. This was critical for two reasons. First, this constraint would result in a tool that could be used with any inpatient administrative data, thus making it useful to most data systems. Second, this would enable national and regional benchmark rates to be provided using HCUP data, since these benchmark rates would need to be calculated using the universe of data available from the States.

2.2 AHRQ Quality Indicator Modules

The work of the UCSF-Stanford EPC resulted in the AHRQ Quality Indicators, which are available as four separate modules:

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• **Prevention Quality Indicators.** These indicators consist of “ambulatory care sensitive conditions,” hospital admissions that evidence suggests could have been avoided through high-quality outpatient care or that reflect conditions that could be less severe, if treated early and appropriately.

• **Inpatient Quality Indicators.** These indicators reflect quality of care inside hospitals and include inpatient mortality; utilization of procedures for which there are questions of overuse, underuse, or misuse; and volume of procedures for which there is evidence that a higher volume of procedures is associated with lower mortality.

• **Patient Safety Indicators.** These indicators also reflect quality of care inside hospitals, but focus on surgical complications and other iatrogenic events.

• **Pediatric Quality Indicators.** This module, available in February, 2006, contains indicators that apply to the special characteristics of the pediatric population.

The core of the Pediatric Quality Indicators (PDIs) is formed by indicators drawn from the original three modules. Some of these indicators were already geared to the pediatric population (for example, PQI 4 – Pediatric Asthma Admission Rate). These indicators are being removed from the original modules.

Others were adapted from indicators that apply to both adult and pediatric populations. These indicators remain in the original module, but will apply only to adult populations.
3.0 Methods of Identifying, Selecting, and Evaluating the Quality Indicators

In developing the new quality indicators, the UCSF-Stanford EPC applied the Institute of Medicine’s widely cited definition of quality care: “the degree to which health services for individuals and populations increase the likelihood of desired health outcomes and are consistent with current professional knowledge.”

They formulated six specific key questions to guide the development process:

- Which indicators are currently in use or described in the literature that could be defined using hospital discharge data?
- What are the quality relationships reported in the literature that could be used to define new indicators using hospital discharge data?
- What evidence exists for indicators not well represented in the original indicators—pediatric conditions, chronic disease, new technologies, and ambulatory care sensitive conditions?
- Which indicators have literature-based evidence to support face validity, precision of measurement, minimum bias, and construct validity of the indicator?
- What risk-adjustment method should be suggested for use with the recommended indicators, given the limits of administrative data and other practical concerns?
- Which indicators perform well on empirical tests of precision of measurement, minimum bias, and construct validity?

As part of this project, the UCSF-Stanford EPC identified quality indicators reported in the literature and used by health care organizations, evaluated the original quality indicators and potential indicators using literature review and empirical methods, incorporated risk adjustment for comparative analysis, and developed new programs that could be employed by users with their own hospital administrative data. This section outlines the steps used to arrive at a final set of quality measures.

3.1 Step 1: Obtain Background Information on QI Use

The project team at the UCSF-Stanford EPC interviewed 33 individuals affiliated with hospital associations, business coalitions, State data groups, Federal agencies, and academia about various topics related to quality measurement, including indicator use, suggested indicators, and other potential contacts. Interviews were tailored to the specific expertise of interviewees. The sample was not intended to be representative of any population; rather, individuals were selected to include QI users and potential users from a broad spectrum of organizations in both the public and private sectors.

Three broad audiences were considered for the quality measures: health care providers and managers, who could use the quality measures to assist in initiatives to improve quality; public health policy makers, who could use the information from indicators to target public health interventions; and health care purchasers, who could use the measures to guide decisions about health policies.

3.2 Step 2: Search the Literature to Identify Potential QIs

The project team performed a structured review of the literature to identify potential indicators. They used Medline to identify the search strategy that returned a test set of known applicable articles in the most concise manner. Using the Medical Subject Heading (MeSH) terms “hospital, statistic, and methods” and “quality indicators” resulted in approximately 2,600 articles published in 1994 or later. After screening

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titles and abstracts for relevancy, the search yielded 181 articles that provided information on potential quality indicators based on administrative data.

Clinicians, health services researchers, and other team members abstracted information from these articles in two stages. In the first stage, preliminary abstraction, they evaluated each of the 181 identified articles for the presence of a defined quality indicator, clinical rationale, and strengths and weaknesses. To qualify for full abstraction, the articles must have explicitly defined a novel quality indicator. Only 27 articles met this criterion. The team collected information on the definition of the quality indicator, validation, and rationale during full abstraction.

In addition, they identified additional potential indicators using the CONQUEST database; the National Library of Healthcare Indicators developed by the Joint Commission on Accreditation of Healthcare Organizations (JCAHO); a list of ORYX-approved indicators provided by JCAHO; and telephone interviews.

3.3 Step 3: Review the Literature to Evaluate the QIs According to Predetermined Criteria

The project team evaluated each potential quality indicator against the following six criteria, which were considered essential for determining the reliability and validity of a quality indicator:

- **Face validity.** An adequate quality indicator must have sound clinical or empirical rationale for its use. It should measure an important aspect of quality that is subject to provider or health care system control.

- **Precision.** An adequate quality indicator should have relatively large variation among providers or areas that is not due to random variation or patient characteristics. This criterion measures the impact of chance on apparent provider or community health system performance.

- **Minimum bias.** The indicator should not be affected by systematic differences in patient case-mix, including disease severity and comorbidity. In cases where such systematic differences exist, an adequate risk adjustment system should be possible using available data.

- **Construct validity.** The indicator should be related to other indicators or measures intended to measure the same or related aspects of quality. In general, better outpatient care (including, in some cases, adherence to specific evidence-based treatment guidelines) can reduce patient complication rates.

- **Fosters real quality improvement.** The indicator should be robust to possible provider manipulation of the system. In other words, the indicator should be insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care.

- **Application.** The indicator should have been used in the past or have high potential for working well with other indicators. Sometimes looking at groups of indicators together is likely to provide a more complete picture of quality.

Based on the initial review, the team identified and evaluated over 200 potential indicators using these criteria. Of this initial set, 45 indicators passed this initial screen and received comprehensive literature and empirical evaluation. In some cases, whether an indicator complemented other promising indicators was a consideration in retaining it, allowing the indicators to provide more depth in specific areas.

For this final set of 45 indicators, the team reviewed an additional 2,000 articles to provide evidence on indicators during the evaluation phase. They searched Medline for articles relating to each of the six areas of evaluation described above. Clinicians and health services researchers reviewed the literature for evidence and prepared a referenced summary description on each indicator.
As part of the review process, the team assessed the link between each indicator and health care quality along the following dimensions:

- **Proxy.** Some indicators do not specifically measure a patient outcome or a process measure of quality. Rather, they measure an aspect of care that is correlated with process measures of quality or patient outcomes. These indicators are best used in conjunction with other indicators measuring similar aspects of clinical care, or when followed with more direct and in-depth investigations of quality.

- **Selection bias.** Selection bias results when a substantial percentage of care for a condition is provided in the outpatient setting, so the subset of inpatient cases may be unrepresentative. In these cases, examination of outpatient care or emergency room data may help reduce selection bias.

- **Information bias.** Quality indicators are based on information available in hospital discharge data sets, but some missing information may actually be important to evaluating the outcomes of hospital care. In these cases, examination of missing information may help to improve indicator performance.

- **Confounding bias.** Patient characteristics may substantially affect performance on a measure and may vary systematically across areas. In these cases, adequate risk adjustment may help to improve indicator performance.

- **Unclear construct validity.** Problems with construct validity include uncertain or poor correlations with widely accepted process measures or with risk-adjusted outcome measures. These indicators would benefit from further research to establish their relationship with quality care.

- **Easily manipulated.** Quality indicators may create perverse incentives to improve performance without actually improving quality. Although very few of these perverse responses have been proven, they are theoretically important and should be monitored to ensure true quality improvement.

- **Unclear benchmark.** For some indicators, the “right rate” has not been established, so comparison with national, regional, or peer group means may be the best benchmark available. Very low PQI rates may flag an underuse problem; that is, providers may fail to hospitalize patients who would benefit from inpatient care. On the other hand, overuse of acute care resources may potentially occur when patients who do not clinically require inpatient care are hospitalized.

### 3.4 Step 4: Perform a Comprehensive Evaluation of Risk Adjustment

The project team identified potential risk-adjustment systems by reviewing the applicable literature and asking the interviewees in step 1 to identify their preferences. Generally, users preferred that the system be (1) open, with published logic; (2) cost-effective, with data collection costs minimized and additional data collection being well justified; (3) designed using a multiple-use coding system, such as those used for reimbursement; and (4) officially recognized by government, hospital groups, or other organizations.

In general, diagnosis-related groups (DRGs) seemed to fit more of the user preference-based criteria than other alternatives. A majority of the users interviewed already used the 3M™ All-Patient Refined DRG® (APR-DRG) system, which has been reported to perform well in predicting resource use and death when compared to other DRG-based systems.

APR-DRGs were used to conduct indicator evaluations to determine the impact of measured differences in patient severity on the relative performance of providers and to provide the basis for implementing APR-DRGs as an optional risk-adjustment system for hospital-level QI measures. The implementation of

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6 Information on the 3M™ APR-DRG system is available at [http://www.3m.com/us/healthcare/his/products/coding/refined_drg.jhtml](http://www.3m.com/us/healthcare/his/products/coding/refined_drg.jhtml).
3.5 Step 5: Evaluate the Indicators Using Empirical Analyses

The project team conducted extensive empirical testing of all potential indicators using the 1995-97 HCUP State Inpatient Databases (SID) and Nationwide Inpatient Sample (NIS) to determine precision, bias, and construct validity. The 1997 SID contains uniform data on inpatient stays in community hospitals for 22 States covering approximately 60% of all U.S. hospital discharges. The NIS is designed to approximate a 20% of U.S. community hospitals and includes all stays in the sampled hospitals. Each year of the NIS contains between 6 million and 7 million records from about 1,000 hospitals. The NIS combines a subset of the SID data, hospital-level variables, and hospital and discharge weights for producing national estimates. The project team conducted tests to examine three things: precision, bias, and construct validity.

Precision. The first step in the analysis involved precision tests to determine the reliability of the indicator for distinguishing real differences in provider performance. For indicators that may be used for quality improvement, it is important to know with what precision, or surety, a measure can be attributed to an actual construct rather than random variation.

For each indicator, the variance can be broken down into three components: variation within a provider (actual differences in performance due to differing patient characteristics), variation among providers (actual differences in performance among providers), and random variation. An ideal indicator would have a substantial amount of the variance explained by between-provider variance, possibly resulting from differences in quality of care, and a minimum amount of random variation. The project team performed four tests of precision to estimate the magnitude of between-provider variance on each indicator:

- Signal standard deviation was used to measure the extent to which performance of the QI varies systematically across hospitals or areas.
- Provider/area variation share was used to calculate the percentage of signal (or true) variance relative to the total variance of the QI.
- Signal-to-noise ratio was used to measure the percentage of the apparent variation in QIs across providers that is truly related to systematic differences across providers and not random variations (noise) from year to year.
- In-sample R-squared was used to identify the incremental benefit of applying multivariate signal extraction methods for identifying additional signal on top of the signal-to-noise ratio.

In general, random variation is most problematic when there are relatively few observations per provider, when adverse outcome rates are relatively low, and when providers have little control over patient outcomes or variation in important processes of care is minimal. If a large number of patient factors that are difficult to observe influence whether or not a patient has an adverse outcome, it may be difficult to separate the “quality signal” from the surrounding noise. Two signal extraction techniques were applied to improve the precision of an indicator:

- Univariate methods were used to estimate the “true” quality signal of an indicator based on information from the specific indicator and 1 year of data.
- Multivariate signal extraction (MSX) methods were used to estimate the “true” quality signal based on information from a set of indicators and multiple years of data. In most cases, MSX methods extracted additional signal, which provided much more precise estimates of true hospital or area quality.
Bias. To determine the sensitivity of potential QIs to bias from differences in patient severity, unadjusted performance measures for specific hospitals were compared with performance measures that had been adjusted for age and gender. All of the PQIs and some of the Inpatient Quality Indicators (IQIs) could only be risk-adjusted for age and sex. The 3M™ APR-DRG System Version 12 with Severity of Illness and Risk of Mortality subclasses was used for risk adjustment of the utilization indicators and the in-hospital mortality indicators, respectively. Five empirical tests were performed to investigate the degree of bias in an indicator:

- Rank correlation coefficient of the area or hospital with (and without) risk adjustment—gives the overall impact of risk adjustment on relative provider or area performance.
- Average absolute value of change relative to mean—highlights the amount of absolute change in performance, without reference to other providers' performance.
- Percentage of highly ranked hospitals that remain in high decile—reports the percentage of hospitals or areas that are in the highest deciles without risk adjustment that remain there after risk adjustment is performed.
- Percentage of lowly ranked hospitals that remain in low decile—reports the percentage of hospitals or areas that are in the lowest deciles without risk adjustment that remain there after risk adjustment is performed.
- Percentage that change more than two deciles—identifies the percentage of hospitals whose relative rank changes by a substantial percentage (more than 20%) with and without risk adjustment.

Construct validity. Construct validity analyses provided information regarding the relatedness or independence of the indicators. If quality indicators do indeed measure quality, then two measures of the same construct would be expected to yield similar results. The team used factor analysis to reveal underlying patterns among large numbers of variables—in this case, to measure the degree of relatedness between indicators. In addition, they analyzed correlation matrices for indicators.
4.0 **Summary Evidence on the Prevention Quality Indicators**

The rigorous evaluations performed by the UCSF-Stanford EPC, based on literature review and empirical testing of indicators, resulted in 16 indicators that reflect ambulatory care sensitive conditions (ACSCs). These ACSCs have been reported and tested in a number of published studies involving consensus processes involving panels of expert physicians, using a range of methodologies and decision criteria. Two sets of ambulatory care sensitive conditions are widely used:

- The set developed by John Billings in conjunction with the United Hospital Fund of New York includes 28 ambulatory care sensitive conditions, identified by a panel of six physicians.\(^7\)
- The set developed by Joel Weissman includes 12 avoidable admissions identified through review of the literature and evaluation by a panel of physicians.\(^8\)

Many of the ACSCs have practice guidelines associated with them, including almost all of the chronic conditions and about half of the acute medical or pediatric conditions. Studies have shown that better outpatient care (including, in some cases, adherence to specific evidence-based treatment guidelines) can reduce patient complication rates of existing disease, including complications leading to hospital admissions. Empirically, most of the hospital admission rates for ACSCs are correlated with each other, suggesting that common underlying factors influence many of the rates.

Five of these 16 PQIs were included in the original HCUP QIs—perforated appendix, low birth weight, pediatric asthma, diabetes short-term complications, and diabetes long-term complications—where they were measured at the hospital level. In contrast, the 16 new indicators were constructed at the community level, defined as a Metropolitan Statistical Area (MSA) or a rural county. For each indicator, lower rates indicate potentially better quality.

4.1 **Version 3.0a PQIs**

A modified version of the process described in Section 3.0 is repeated on an annual basis when the PQIs are evaluated and new indicators are considered. With this release two of the original 16 indicators dealing with pediatric asthma and pediatric gastroenteritis have been moved to the Pediatric Quality Indicators (PDI) module.

Mew micropolitan statistical areas and updated metropolitan statistical areas were established by the federal Office of Management and Budget (OMB) circular 03-04 (last revised December 4, 2005). To reflect these changes, all PQI documentation now refers to Metro Area instead of MSA. The SAS and SPSS software allows users to specify stratification by county level with U.S. Census FIPS or modified FIPS, or by Metro Area with OMB 1999 or OMB 2003 definition. The AHRQ QI Windows Application allows users to generate reports stratified by all four of these, as well as by State. See Appendix A for links to additional information.

Table 1 summarizes the results of the literature review and empirical evaluations on the PQIs. It lists each indicator, provides its definition, rates its empirical performance, recommends a risk adjustment strategy, and summarizes important caveats identified from the literature review.

Rating of performance on empirical evaluations, as described in step 5 above, ranged from 0 to 26. (The average score for the 16 original PQIs is 14.6.) The scores were intended as a guide for summarizing the performance of each indicator on four empirical tests of precision (signal variance, area-level share,  

\(^8\)Weissman, JS, Gatsonis C, Epstein AM. Rates of avoidable hospitalization by insurance status in Massachusetts and Maryland. JAMA 1992;268(17):2388-94.
signal ratio, and R-squared) and five tests of minimum bias (rank correlation, top and bottom decile movement, absolute change, and change over two deciles), as described in the previous section.

The magnitude of the scores, shown in the Empirical Rating column, provides an indication of the relative rankings of the indicators. These scores were based on indicator performance after risk-adjustment and smoothing; that is, they represent the “best estimate” of the indicator’s true value after accounting for case-mix and reliability. The score for each individual test is an ordinal ranking (e.g., very high, high, moderate, and low). The final summary score was derived by assigning a weight to each ranking (e.g., 3, 2, 1, 0) and summing across these nine individual tests. Higher scores indicate better performance on the empirical tests.

The Literature Review Findings column summarizes evidence specific to each potential concern on the link between the PQIs and quality of care, as described in step 3 above. A question mark (?) indicates that the concern is theoretical or suggested, but no specific evidence was found in the literature. A check mark (✓) indicates that the concern has been demonstrated in the literature.

A complete description of each PQI is included later in the guide in Section 5.0, “Detailed Evidence for Prevention Quality Indicators” that starts on page 17, and in the document Prevention Quality Indicators Technical Specifications. (See Appendix A.)

### Table 1. Prevention Quality Indicators

<table>
<thead>
<tr>
<th>Indicator Name (Number)</th>
<th>Description</th>
<th>Risk Adjustment Incorporated</th>
<th>Empirical Performance</th>
<th>Literature Review Findings</th>
</tr>
</thead>
<tbody>
<tr>
<td>Diabetes Short-term Complication Admission Rate (PQI 1)</td>
<td>Number of admissions for diabetes short-term complications per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 48.2 Area SD: 41.8 Pop. Rate: 51.1 Rating: 14</td>
<td>? Proxy ? Confounding bias</td>
</tr>
<tr>
<td>Perforated Appendix Admission Rate (PQI 2)</td>
<td>Number of admissions for perforated appendix as a share of all admissions for appendicitis within an area.</td>
<td>Age and sex.</td>
<td>Area Rate: 34.9 per 100 Area SD: 18.3 Pop. Rate: 30.6 per 100 Rating: 17</td>
<td>? Proxy</td>
</tr>
<tr>
<td>Diabetes Long-term Complication Admission Rate (PQI 3)</td>
<td>Number of admissions for long-term diabetes per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 116.0 Area SD: 74.0 Pop. Rate: 115.4 Rating: 11</td>
<td>? Proxy ? Confounding bias ? Easily manipulated ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Chronic Obstructive Pulmonary Disease Admission Rate (PQI 5)</td>
<td>Number of admissions for COPD per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 349.1 Area SD: 290.5 Pop. Rate: 244.2 Rating: 17</td>
<td>? Proxy ? Confounding bias ? Easily manipulated ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Hypertension Admission Rate (PQI 7)</td>
<td>Number of admissions for hypertension per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 51.2 Area SD: 53.5 Pop. Rate: 45.1 Rating: 14</td>
<td>? Proxy ? Easily manipulated ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Indicator Name (Number)</td>
<td>Description</td>
<td>Risk Adjustment Incorporated</td>
<td>Empirical Performancea</td>
<td>Literature Review Findingsb</td>
</tr>
<tr>
<td>------------------------</td>
<td>-------------</td>
<td>-----------------------------</td>
<td>------------------------</td>
<td>---------------------------</td>
</tr>
<tr>
<td>Congestive Heart Failure Admission Rate (PQI 8)</td>
<td>Number of admissions for CHF per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 525.1  Area SD: 267.6  Pop. Rate: 468.4  Rating: 14</td>
<td>? Proxy  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Low Birth Weight Rate (PQI 9)</td>
<td>Number of low birth weight births as a share of all births in an area.</td>
<td>Not risk adjusted.</td>
<td>Area Rate: 5.7 per 100  Area SD: 3.6  Pop. Rate: 5.8 per 100  Rating: 11c out of 16d</td>
<td>? Proxy  ? Confounding bias  ✓ Unclear construct</td>
</tr>
<tr>
<td>Dehydration Admission Rate (PQI 10)</td>
<td>Number of admissions for dehydration per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 167.7  Area SD: 128.1  Pop. Rate: 127.7  Rating: 14</td>
<td>? Proxy  ? Unclear construct  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Bacterial Pneumonia Admission Rate (PQI 11)</td>
<td>Number of admissions for bacterial pneumonia per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 598.9  Area SD: 352.5  Pop. Rate: 420.7  Rating: 17</td>
<td>? Proxy  ? Unclear construct  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Urinary Tract Infection Admission Rate (PQI 12)</td>
<td>Number of admissions for urinary infection per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 202.0  Area SD: 133.1  Pop. Rate: 170.2  Rating: 11</td>
<td>? Proxy  ? Unclear construct  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Angina without Procedure Admission Rate (PQI 13)</td>
<td>Number of admissions for angina without procedure per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 68.5  Area SD: 68.6  Pop. Rate: 45.9  Rating: 19</td>
<td>? Proxy  ? Unclear construct  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Uncontrolled Diabetes Admission Ratea (PQI 14)</td>
<td>Number of admissions for uncontrolled diabetes per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 25.4  Area SD: 30.6  Pop. Rate: 21.0  Rating: 14</td>
<td>? Proxy  ? Confounding bias  ? Easily manipulated</td>
</tr>
<tr>
<td>Adult Asthma Admission Rate (PQI 15)</td>
<td>Number of admissions for asthma in adults per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 116.7  Area SD: 87.4  Pop. Rate: 125.1  Rating: 16</td>
<td>? Proxy  ? Easily manipulated  ✓ Unclear benchmark</td>
</tr>
<tr>
<td>Rate of Lower-extremity Amputation Among Patients with Diabetes (PQI 16)</td>
<td>Number of admissions for lower-extremity amputation among patients with diabetes per 100,000 population.</td>
<td>Age and sex.</td>
<td>Area Rate: 37.7  Area SD: 29.5  Pop. Rate: 36.6  Rating: 10c</td>
<td>? Proxy  ? Unclear construct</td>
</tr>
</tbody>
</table>

a Higher scores in the Empirical Performance column indicate better performance on the nine empirical tests. Unadjusted means and standard deviations (SD) were calculated using the 2003 SID from 38 states. The area rates are average area rates and area standard deviation based on 2,553 geographic areas (counties) in the 2003 SID. The population rate is based...
AHRQ Quality Indicators Web Site: http://www.qualityindicators.ahrq.gov

The software provides the option to generate condition-specific rates (e.g., using the number of diabetics in the denominator) by state and age. Table 2 provides the Empirical Performance rates for the four diabetes-related PQIs, expressed per 1,000.

Table 2. Diabetes-related Prevention Quality Indicators

<table>
<thead>
<tr>
<th>Indicator Name (Number)</th>
<th>Description</th>
<th>Risk Adjustment Incorporated</th>
<th>Empirical Performance</th>
<th>Literature Review Findings</th>
</tr>
</thead>
<tbody>
<tr>
<td>Diabetes Short-term Complication Admission Rate (PQI 1)</td>
<td>Number of admissions for diabetes short-term complications per 1,000 diabetic.</td>
<td>N/A</td>
<td>Area Rate: 7.5 &lt;br&gt; Area SD: 1.3 &lt;br&gt; Pop. Rate: 7.5 &lt;br&gt; Rating: 14</td>
<td>? – Proxy ? – Confounding bias</td>
</tr>
<tr>
<td>Diabetes Long-term Complication Admission Rate (PQI 3)</td>
<td>Number of admissions for long-term diabetes per 1,000 diabetic.</td>
<td>N/A</td>
<td>Area Rate: 16.2 &lt;br&gt; Area SD: 2.4 &lt;br&gt; Pop. Rate: 16.8 &lt;br&gt; Rating: 11</td>
<td>? – Proxy ? – Confounding bias ? – Easily manipulated ▼ – Unclear benchmark</td>
</tr>
<tr>
<td>Uncontrolled Diabetes Admission Rate* (PQI 14)</td>
<td>Number of admissions for uncontrolled diabetes per 1,000 diabetic.</td>
<td>N/A</td>
<td>Area Rate: 2.6 &lt;br&gt; Area SD: 1.3 &lt;br&gt; Pop. Rate: 3.1 &lt;br&gt; Rating: 14</td>
<td>? – Proxy ? – Confounding bias ? – Easily manipulated</td>
</tr>
<tr>
<td>Rate of Lower-extremity Amputation Among Patients with Diabetes (PQI 16)</td>
<td>Number of admissions for lower-extremity amputation among patients with diabetes per 1,000 diabetic.</td>
<td>N/A</td>
<td>Area Rate: 5.2 &lt;br&gt; Area SD: 0.9 &lt;br&gt; Pop. Rate: 5.3 &lt;br&gt; Rating: 10°</td>
<td>? – Proxy ? – Unclear construct</td>
</tr>
</tbody>
</table>

* Based on 38 states in 2003 SID.

4.2 Strengths and Limitations in Using the PQIs

The PQIs represent the current state of the art in assessing quality of health services in local communities using inpatient discharge data. These indicators measure the outcomes of preventive care for both acute illness and chronic conditions, reflecting two important components of the quality of preventive care—effectiveness and timeliness. For example, with effective drug therapy in the outpatient setting, hospital admissions for hypertension can be prevented. Likewise, accurate diagnosis and timely access to surgical treatment will help reduce the incidence of perforated appendix. The PQIs are thus valuable tools...
for identifying potential quality problems in outpatient care that help to set the direction for more in-depth investigation. Because the PQIs are based on readily available data—hospital discharge abstracts—resource requirements are minimal. With uniform definitions and standardized programs, the PQIs will allow comparisons across States, regions, and local communities over time.

Despite the unique strengths of the PQIs, there are several issues that should be considered when using these indicators. First, for some PQIs, differences in socioeconomic status have been shown to explain a substantial part—perhaps most—of the variation in PQI rates across areas. The complexity of the relationship between socioeconomic status and PQI rates makes it difficult to delineate how much of the observed relationships are due to true access to care difficulties in potentially underserved populations, or due to other patient characteristics, unrelated to quality of care, that vary systematically by socioeconomic status. For some of the indicators, patient preferences and hospital capabilities for inpatient or outpatient care might explain variations in hospitalizations. In addition, environmental conditions that are not under the direct control of the health care system can substantially influence some of the PQIs. For example, the COPD and asthma admission rates are likely to be higher in areas with poorer air quality.

Second, the evidence related to potentially avoidable hospital admissions is limited for each indicator, because many of the indicators have been developed as parts of sets. Only five studies have attempted to validate individual indicators rather than whole measure sets.\(^9\)\(^{10}\)\(^{11}\)\(^{12}\)\(^{13}\) A limitation of this literature is that relatively little is known about which components represent the strongest measures of access and quality. Most of the five papers that did report on individual indicators also used a single variable, such as median area-specific income or rural residence, for construct validation. All but one of these papers\(^10\) included adjustment only for demographic factors (e.g., age, sex, and race).

Third, despite the relationships demonstrated at the patient level between higher quality ambulatory care and lower rates of hospital admission, few studies have directly addressed the question of whether effective treatments in outpatient settings would reduce the overall incidence of hospitalizations. The extent to which the reporting of admission rates for ambulatory care sensitive conditions may lead to changes in ambulatory practices and admission rates also is unknown. Providers may admit patients who do not clinically require inpatient care or they may do the opposite—fail to hospitalize patients who would benefit from inpatient care.

4.3 Questions for Future Work

The limitations discussed above suggest some directions for future work on development and use of the PQIs. Additional data and linkages could provide insights into the underlying causes of hospitalization for these conditions and could facilitate the exploration of potential interventions to prevent such events.

- Studies examining health and risk behaviors in a population could illuminate patient factors associated with the incidence of ambulatory care sensitive conditions.
- Examining environmental data, such as air pollution levels, could provide insight into factors outside the direct control of the health care system that are associated with hospitalization for such conditions.
- Exploring differences in disease prevalence in specific areas could help to discern whether variations in hospitalization rates can be attributed to differences in disease burden across communities that would exist even with optimum preventive care.

• Studies could examine the relationship between rural-urban location and distance to health care resources and hospital admission for ambulatory care sensitive conditions. Such studies would require information on patients' residence such as patient ZIP codes.

• Linkages with data on local medical resources could help to illuminate the relationship between hospitalization for ACSCs and the supply of medical services and resources, such as the number of primary care and specialty physicians in a community or the supply of hospital beds. For example, the Dartmouth Atlas provides analyses for the Medicare population that suggest that the supply of hospital beds in a community is linked to ambulatory care sensitive admissions, but reported no relationship with local physician supply.\(^\text{14}\)

• Physician office data and outpatient clinic data may provide important information regarding care prior to hospital admission. Outpatient data would enable analyses that examine the processes of care that can prevent hospitalizations due to these conditions.

• Combining inpatient data with emergency department data would support the construction of a more complete picture of quality of care related to ambulatory care sensitive conditions. Some of these conditions are seen in emergency departments without being admitted for inpatient care. This is particularly relevant for the uninsured or underinsured who are more likely to use emergency departments as a routine source of care.

5.0 Detailed Evidence for Prevention Quality Indicators

This section provides an abbreviated presentation of the details of the literature review and the empirical evaluation for each PQI, including:

- The relationship between the indicator and quality of health care services
- A suggested benchmark or comparison
- The definition of each indicator
- The outcome of interest (or numerator)
- The population at risk (or denominator)
- The results of the empirical testing

Empirical testing rated the statistical performance of each indicator, as described in step 5 in the previous section. Scores ranged from 0 to 26 (mean for the 16 original PQIs = 14.6), except for low birth weight for which bias was not tested because adequate risk adjustment was not available. The scores are intended as a guide for summarizing the performance of each indicator on four empirical tests of precision (signal variance, area-level share, signal ratio, and R-squared) and five tests of minimum bias (rank correlation, top and bottom decile movement, absolute change, and change over two deciles), as described in the previous section. Raw unadjusted rates and SD are calculated using 2003 SID from 38 states. These rates are population rates based on all eligible discharges, as opposed to the average area rates reported in Table 1.

The magnitude of the scores, shown under Empirical Rating, provides an indication of the relative rankings of the indicators. These scores were based on indicator performance after risk-adjustment and smoothing, that is, they represent the “best estimate” of the indicator’s true value after accounting for case-mix and reliability. The score for each individual test is an ordinal ranking (e.g., very high, high, moderate, and low). The final summary score was derived by assigning a weight to each ranking (e.g., 3, 2, 1, 0) and summing across these nine individual tests. Higher scores indicate better performance on the empirical tests. The two-page descriptions for each indicator also include a discussion of the summary of evidence, the limitations on using each indicator, and details on:

- Face validity – Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?
- Precision – Is there a substantial amount of provider or community level variation that is not attributable to random variation?
- Minimum bias – Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

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15 The state data organizations that participated in the 2003 HCUP SID: Arizona Department of Health Services; California Office of Statewide Health Planning & Development; Colorado Health & Hospital Association; Connecticut - Chime, Inc.; Florida Agency for Health Care Administration; Georgia: An Association of Hospitals & Health Systems; Hawaii Health Information Corporation; Illinois Health Care Cost Containment Council; Indiana Hospital & Health Association; Iowa Hospital Association; Kansas Hospital Association; Kentucky Department for Public Health; Maine Health Data Organization; Maryland Health Services Cost Review; Massachusetts Division of Health Care Finance and Policy; Michigan Health & Hospital Association; Minnesota Hospital Association; Missouri Hospital Industry Data Institute; Nebraska Hospital Association; Nevada Department of Human Resources; New Hampshire Department of Health & Human Services; New Jersey Department of Health & Senior Services; New York State Department of Health; North Carolina Department of Health and Human Services; Ohio Hospital Association; Oregon Association of Hospitals & Health Systems; Pennsylvania Health Care Cost Containment Council; Rhode Island Department of Health; South Carolina State Budget & Control Board; South Dakota Association of Healthcare Organizations; Tennessee Hospital Association; Texas Health Care Information Council; Utah Department of Health; Vermont Association of Hospitals and Health Systems; Virginia Health Information; Washington State Department of Health; West Virginia Health Care Authority; Wisconsin Department of Health & Family Services.
• Construct validity – Does the indicator perform well in identifying true (or actual) quality-of-care problems?

• Fosters true quality improvement – Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

• Prior use – Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

Summary of Evidence Reported for All or Multiple PQIs

The literature review of the evidence related to potentially avoidable hospital admissions is limited for each indicator because many of the individual indicators have been developed as parts of sets. This section provides a summary of the general evidence reviewed applicable to all PQIs.

• **Precision.** The precision of avoidable hospitalization rates is likely to depend on the size of the denominator.

• **Minimum bias.** Previous studies have documented several characteristics that are associated with either the risk of an avoidable hospitalization (at the individual level) or the avoidable hospitalization rate (at the area level), including prevalence of the condition, race, socioeconomic status (SES), chronic disease and health of the population. These characteristics may be confounding factors, but also might be measuring subtle aspects of access to care.

• **Construct validity.** Most previous studies have assessed the validity of an entire set of avoidable hospital conditions, rather than each condition alone, and have used SES as a marker of access to care. These studies have repeatedly shown strong correlations between household income and avoidable hospitalizations, both at the individual level and the area level. At the zip code level, income alone explains 51-84% of the variability in ACS admission rates across 15 metropolitan areas in the US. This association is substantially weaker among persons 65 or more years of age, as one would expect if it is driven by access to care rather than underlying social factors. Avoidable hospitalization rates are higher among uninsured or Medicaid-enrolled persons than among privately insured persons, even after adjustment for race and income.

Fewer studies have tested true measures of access to care. In the best of these studies, Bindman and colleagues showed that self-reported “difficulty in receiving medical care when needed” explained 50% of the variability in hospitalization rates for 5 chronic medical conditions (asthma, CHF, COPD, diabetes, and hypertension). Adjustment for condition prevalence, propensity to seek care, physician admitting style, and ecological measures of income, education, insurance, race, and gender, had little effect on the association. Having a regular source of care, and primary care physician/population ratios, were also independently associated with avoidable hospitalization rates, when substituted for self-reported access.

These relationships did not hold in two separate studies of rural zip codes, suggesting that avoidable hospitalization rates are invalid indicators of access in rural areas.\textsuperscript{24,25}

In other studies, the physician/population ratio for family and general physicians has been more strongly associated with avoidable hospitalization rates than measures that include internists, pediatricians, or all physicians.\textsuperscript{26-27} In studies of Medicaid populations, provider continuity in ambulatory care\textsuperscript{28} and usual care received from a community health center\textsuperscript{29} were associated with lower avoidable hospitalization rates, and not having a primary care physician was associated with higher rates of avoidable hospitalization.\textsuperscript{30} However, having a regular source of care (for more than 50\% of physician office visits) was not associated with lower avoidable hospitalization rates.\textsuperscript{31}

Several studies of Medicare beneficiaries have shown weak and inconsistent associations between access indicators and avoidable hospitalization rates. For example, persons in the Medicare Current Beneficiary Survey who reported problems obtaining health care, or lived in a health professional shortage area, were not at increased risk of preventable hospitalization.\textsuperscript{17} Instead, their risk was heavily influenced by clinical factors. However, beneficiaries in fair or poor health reportedly were at increased risk if they lived in a primary care shortage area.\textsuperscript{32} An area-level analysis based on Medicare claims suggests that the association between admission rates and physician/population ratios is limited to the 10\% of health care service areas with the most severe shortage of physicians.\textsuperscript{33}

A full report on the literature review and empirical evaluation can be found in \textit{Refinement of the HCUP Quality Indicators} by the UCSF-Stanford EPC, Detailed coding information for each PQI is provided in the document \textit{Prevention Quality Indicators Technical Specifications}. See Appendix A for links to these and other documents.

\begin{thebibliography}{9}
\bibitem{27} Epstein A. The role of the medical market in preventable hospitalizations. Abstract Book/Association of Health Services Research 1998;15(316-7).
\bibitem{31} Gill JM. Can hospitalizations be avoided by having a regular source of care? Fam Med 1997;29(3):166-71.
\end{thebibliography}
5.1 Diabetes Short-term Complications Admission Rate (PQI 1)

Short-term complications of diabetes mellitus include diabetic ketoacidosis, hyperosmolarity, and coma. These life-threatening emergencies arise when a patient experiences an excess of glucose (hyperglycemia) or insulin (hypoglycemia).

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment and adherence to care may reduce the incidence of diabetic short-term complications, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for diabetic short-term complications per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for diabetes short-term complications (ketoacidosis, hyperosmolarity, coma).</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 51.1 per 100,000 population</td>
</tr>
<tr>
<td></td>
<td>Rating: 14</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for diabetes short-term complications is a PQI that would be of most interest to comprehensive health care delivery systems. Short-term diabetic emergencies arise from the imbalance of glucose and insulin, which can result from deviations in proper care, misadministration of insulin, or failure to follow a proper diet.

Although risk adjustment with age and sex does not impact the relative or absolute performance of areas, this indicator should be risk-adjusted. Some areas may have higher rates of diabetes as a result of racial composition and systematic differences in other risk factors.

Areas with high rates of diabetic emergencies may want to examine education practices, access to care, and other potential causes of non-compliance when interpreting this indicator. Also, areas may consider examining the rates of hyperglycemic versus hypoglycemic events when interpreting this indicator.

Limitations on Use

As a PQI, short-term diabetes complication rate is not a measure of hospital quality, but rather one measure of outpatient and other health care. Rates of diabetes may vary systematically by area, creating bias for this indicator. Examination of both inpatient and outpatient data may provide a more complete picture of diabetes care.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

High-quality outpatient management of patients with diabetes has been shown to lead to reductions in almost all types of serious avoidable hospitalizations. However, tight control may be associated with more episodes of hypoglycemia, which leads to more admissions.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Based on empirical evidence, this indicator is moderately precise, with a raw area level rate of 36 per 100,000 population and a standard deviation of 24.6.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance)
rather than random variation) is moderate, at 51.7%, indicating that some of the observed differences in age-sex adjusted rates do not represent true differences in area performance. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Minorities have higher rates of diabetes, and higher hospitalization rates may result in areas with higher minority concentrations. Empirical results show that area rankings and absolute performance are not affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Studies of precipitating events of admission for diabetic emergencies often rely on self-report, which may be a biased measurement in and of itself. The results of one study showed that over 60% of patients with known and treated diabetes had made an error in insulin administration or had omitted insulin. In a potentially underserved population of urban African-Americans, two-thirds of admissions were due to cessation of insulin therapy—over half of the time for financial or other difficulties obtaining insulin.

Bindman reported that an area’s self-rated access to care report explained 46% of the variance in admissions for diabetes, although the analysis was not restricted to diabetic emergencies. Weissman found that uninsured patients had more than twice the risk of admission for diabetic ketoacidosis and coma than privately insured patients.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Because diabetic emergencies are potentially life-threatening, hospitals are unlikely to fail to admit patients requiring hospitalization.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

Admission for diabetic emergencies was included in both Billings and Weissman’s sets of avoidable hospitalization measures. This indicator, defined as a provider-level indicator, was an original HCUP QI.

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5.2 Perforated Appendix Admission Rate (PQI 2)

Perforated appendix may occur when appropriate treatment for acute appendicitis is delayed for a number of reasons, including problems with access to care, failure by the patient to interpret symptoms as important, and misdiagnosis and other delays in obtaining surgery.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Timely diagnosis and treatment may reduce the incidence of perforated appendix, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for perforated appendix per 100 admissions for appendicitis within Metro Area or county.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM diagnosis code for perforation or abscess of appendix in any field.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Discharges with diagnosis code for appendicitis in any field within Metro Area or county.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 30.6 per 100 eligible discharges</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for perforated appendix is a PQI that would be of most interest to comprehensive health care delivery systems. With prompt and appropriate care, acute appendicitis should not progress to perforation or rupture. Rates for perforated appendix are higher in the uninsured or underinsured in both adult and pediatric populations, which may be caused by patients failing to seek appropriate care, difficulty in accessing care, or misdiagnoses and poor quality care.

Perforated appendix rates vary systematically by race, although the cause is unknown. Areas with high rates of perforated appendix may want to target points of intervention by using chart reviews and other supplemental data to investigate the reasons for delay in receiving surgery. Hospital contributions to the overall area rate may be particularly useful for this indicator, because misdiagnoses and other delays in receiving surgery in an emergency room may contribute substantially to the rate.

Limitations on Use

As a PQI, admission for perforated appendix is not a measure of hospital quality, but rather one measure of outpatient and other health care.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Perforated appendix results from delay in surgery, potentially reflecting problems in access to ambulatory care, misdiagnosis, and other delays in obtaining surgery.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Perforated appendix occurs in one-fourth to one-third of hospitalized acute appendicitis patients. Based on empirical evidence, this indicator is precise, with a raw area level rate of 33.3% and a substantial standard deviation of 14.4%.

Relative to other indicators, a higher percentage of the variation occurs at the area level rather than the discharge level. However, the signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Higher rates of perforated appendix have been noted in males, patients with mental illness or substance abuse disorders, people with diabetes, and blacks, as well as in children under the age of 4 (although appendicitis is rare in this age group).

Some of the observed variation in performance is due to systematic differences in patient characteristics. No evidence exists in the literature that clinical characteristics that would vary systematically increase the likelihood of perforated appendix. Therefore, this indicator is unlikely to be clinically biased. Empirical results show that area rankings and absolute performance are not affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Braveman et al. found that the rate of perforated appendix was 50% higher for patients with no insurance or Medicaid than HMO-covered patients, and 20% higher for patients with private fee-for-service insurance. A follow-up study by Blumberg et al. concluded that the high rate of perforated appendix in the black population at an HMO may be explained by delay in seeking care, rather than differences in the quality of health care.

Use of this quality indicator might lead to more performance of appendectomies in cases of questionable symptoms, in addition to reducing the occurrence of rupture.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

Perforated appendix was included in the original HCUP QI indicator set, as well as in Weissman’s set of avoidable hospitalizations.

Based on empirical results, areas with high rates of perforated appendix admissions tend to have lower rates of admissions for other ACSCs.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

41Braveman et al., 1994.

5.3 Diabetes Long-term Complications Admission Rate (PQI 3)

Long-term complications of diabetes mellitus include renal, eye, neurological, and circulatory disorders. Long-term complications occur at some time in the majority of patients with diabetes to some degree.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment and adherence to care may reduce the incidence of diabetic long-term complications, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for diabetic long-term complications per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for long-term complications of diabetes (renal, eye, neurological, circulatory, or complications not otherwise specified).</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 115.4 per 100,000 population</td>
</tr>
<tr>
<td></td>
<td>Rating: 11</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for diabetes long-term complications is a PQI that would be of most interest to comprehensive health care delivery systems. Long-term diabetes complications are thought to arise from sustained long-term poor control of diabetes. Intensive treatment programs have been shown to decrease the incidence of long-term complications in both Type 1 and Type 2 diabetes.

Sociodemographic characteristics of the population, such as race, may bias the indicator, since Native Americans and Hispanic Americans have higher rates of diabetes and poorer glycemic control. The importance of these factors as they relate to admission rates is unknown. Risk adjustment for observable characteristics, such as racial composition of the population, is recommended.

It is unclear whether poor glycemic control arises from poor quality medical care, non-compliance of patients, lack of education, or access to care problems. Areas with high rates may wish to examine these factors when interpreting this indicator.

Limitations on Use

As a PQI, diabetes long-term complication rate is not a measure of hospital quality, but rather one measure of outpatient and other health care. Rates of diabetes may vary systematically by area, creating bias for this indicator. Examination of both inpatient and outpatient data may provide a more complete picture of diabetes care.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Several observational studies have linked improved glycemic control to substantially lower risks of developing complications in both Type 1 and Type 2 diabetes. Given that appropriate adherence to therapy and consistent monitoring of glycemic control help to prevent complications, high-quality outpatient care should lower long-term complication rates. However, adherence to guidelines aimed at reducing complications (including eye and foot examinations and diabetic education) has been

described as modest, with only one-third of patients receiving all essential services.

**Precision:** Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Diabetes affects a large number of people, as do diabetic complications. However, few studies have documented hospitalization rates for diabetic complications and the extent to which they vary across areas. Based on empirical evidence, this indicator is moderately precise, with a raw area level rate of 80.8 per 100,000 population and a standard deviation of 58.1.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is high, at 75.6%, indicating that the observed differences in age-sex adjusted rates likely represent true differences across areas.

**Minimum bias:** Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Rates of diabetes are higher in black, Hispanic, and especially Native American populations than in other ethnic groups. Hyperglycemia appears to be particularly frequent among Hispanic and Native American populations. The duration of diabetes is positively associated with the development of complications. Empirical results show that area rankings and absolute performance are moderately affected by age-sex risk adjustment.

**Construct validity:** Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Compliance of physicians and patients is essential to achieve good outcomes, and it seems likely that problems with both access to and quality of care, as well as patient compliance, may contribute to the occurrence of complications.

Based on empirical results, areas with high rates of diabetes long-term complications also tend to have high rates of admission for other ACSCs.

**Fosters true quality improvement:** Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Providers may decrease admission rates by failing to hospitalize patients who would truly benefit from inpatient care. No published evidence indicates that worse health outcomes are associated with reduced hospitalization rates for long-term complications of diabetes.

**Prior use:** Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator, defined as a hospital-level indicator, is an original HCUP QI.

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5.4 Chronic Obstructive Pulmonary Disease Admission Rate (PQI 5)

Chronic obstructive pulmonary disease (COPD) comprises three primary diseases that cause respiratory dysfunction—asthma, emphysema, and chronic bronchitis—each with distinct etiologies, treatments, and outcomes. This indicator examines emphysema and bronchitis; asthma is discussed separately for children and adults.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for COPD, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for COPD per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for COPD.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
</tbody>
</table>
| Empirical Results and Rating | Rate (2003): 244.2 per 100,000 population
|                          | Rating: 17                                                                                               |

Summary of Evidence

Hospital admission for COPD is a PQI that would be of most interest to comprehensive health care delivery systems. COPD can often be controlled in an outpatient setting. Areas may wish to use chart reviews to understand more clearly whether admissions are a result of poor quality care or other problems.

This indicator is measured with high precision, and the observed variance likely reflects true differences across areas. Risk adjustment for age and sex appears to most affect the areas with the highest rates. Several factors that are likely to vary by area may influence the progression of the disease, including smoking and socioeconomic status. Risk adjustment for observable characteristics is recommended.

Areas may wish to identify hospitals that contribute the most to the overall area rate for this indicator. The patient populations served by these hospitals may be a starting point for interventions.

Limitations on Use

As a PQI, COPD is not a measure of hospital quality, but rather one measure of outpatient and other health care. This indicator has unclear construct validity, because it has not been validated except as part of a set of indicators. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting. Some COPD care takes place in emergency rooms, so combining inpatient and emergency room data may give a more accurate picture.

Details

*Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?*

Admissions for COPD include exacerbations of COPD, respiratory failure, and (rarely) lung volume reduction surgery or lung transplantation. Practice guidelines for COPD have been developed and published over the last decade. With appropriate outpatient treatment and compliance, hospitalizations for the exacerbations of COPD and decline in lung function should be minimized.

*Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?*

COPD accounts for a substantial number of hospital admissions, suggesting that the

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indicator is reasonably precise. Based on empirical evidence, this indicator is very precise, with a raw area level rate of 324.0 per 100,000 population and a standard deviation of 203.8.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is very high, at 93.4%, indicating that the differences in age-sex adjusted rates likely represent true differences across areas.

Minimal bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Factors that have been associated with increased admissions for COPD include disease severity, smoking status, age, and socioeconomic status, which are candidates for risk adjustment. Empirical results show that area rankings and absolute performance are somewhat affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Bindman et al. reported that self-reported access to care explained 27% of the variation in COPD hospitalization rates at the ZIP code cluster level. Millman et al. found that low-income ZIP codes had 5.8 times more COPD hospitalizations per capita than high-income ZIP codes. Physician adherence to practice guidelines and patient compliance also influence the effectiveness of therapy.

Based on empirical results, areas with high rates of COPD admissions also tend to have high rates of admissions for other ACSCs.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

One study found that higher rates of COPD admission may in part reflect improvements in access to care, which results in more detection of significant respiratory impairment in the community. A decline in COPD admission rates may simply reflect a reverse change in coding practices.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was originally developed by Billings et al. in conjunction with the United Hospital Fund of New York. It was subsequently adopted by the Institute of Medicine and has been widely used in studies of avoidable hospitalizations.


5.5 Hypertension Admission Rate (PQI 7)

Hypertension is a chronic condition that is often controllable in an outpatient setting with appropriate use of drug therapy.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for hypertension, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for hypertension per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for hypertension.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude discharges with specified cardiac procedure codes in any field, patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 45.1 per 100,000 population Rating: 14</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for hypertension is a PQI that would be of most interest to comprehensive health care delivery systems. Little evidence exists regarding the validity of this indicator, although one study did relate admission rates to access to care problems. This indicator is measured with adequate precision, but some of the variance in age-sex adjusted rates does not reflect true differences in area performance. Adjustment for age-sex is recommended.

Areas may wish to identify hospitals that contribute the most to the overall area rate for this indicator. The patient populations served by these hospitals may be a starting point for interventions.

Limitations on Use

As a PQI, hypertension is not a measure of hospital quality, but rather one measure of outpatient and other health care. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting.

Details

*Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?*

Hypertension is often controllable in an outpatient setting with appropriate use of drug therapy.

*Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?*

Although hypertension is a common condition, hospitalizations for complications of hypertension are relatively uncommon. One study noted that hypertension accounted for only 0.5% of total admissions for ACSCs.55

Based on empirical evidence, this indicator is moderately precise, with a raw area level rate of 37.1 per 100,000 population and a substantial standard deviation of 32.2. The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is moderate, at 69.9%, indicating that some of the observed differences in age-sex adjusted rates likely do not represent true differences in area performance.

*Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?*

Little evidence exists on potential biases for this indicator. The age structure of the population may possibly affect admission rates for this condition. Weissman et al. reported a reduction of 100% in relative risk for Medicaid patients when adjusting for age and sex. No evidence was found on the effects of comorbidities such as obesity or other risk factors that may vary systematically by area on admission rates for hypertension complications in the area. Empirical results show that age-sex adjustment affects the ranking of those areas in the highest decile.

*Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?*

Bindman et al. found that an area’s self-rated access to care explained 22% of admissions for hypertension. Weissman et al. found that uninsured patients had a relative risk of admission for hypertension of 2.38 in Massachusetts after adjustment for age and sex, while Maryland had a corresponding relative risk of 1.93. Millman et al. reported that low-income ZIP codes had 7.6 times more hypertension hospitalizations per capita than high-income ZIP codes.

*Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?*

Little evidence exists on the impact of this quality improvement measure on the delivery of outpatient care for hypertension. There is no published evidence of worse health outcomes in association with reduced hospitalization rates for hypertension. Such an effect seems implausible, given that only the most serious episodes of accelerated or malignant hypertension are treated on an inpatient basis.


Prior use: Has the indicator been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was included originally developed by Billings et al. in conjunction with the United Hospital Fund of New York. It was subsequently adopted by the Institute of Medicine and has been widely used in a variety of studies of avoidable or preventable hospitalizations. This indicator was also included in Weissman’s set of avoidable hospitalizations.

5.6 Congestive Heart Failure Admission Rate (PQI 8)

Congestive heart failure (CHF) can be controlled in an outpatient setting for the most part; however, the disease is a chronic progressive disorder for which some hospitalizations are appropriate.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for CHF, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for CHF per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for CHF.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients discharged with specified cardiac procedure codes in any field, patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 468.4 per 100,000 population</td>
</tr>
<tr>
<td></td>
<td>Rating: 14</td>
</tr>
</tbody>
</table>

Summary of Evidence

Congestive heart failure is a PQI that would be of most interest to comprehensive health care delivery systems. This indicator is measured with high precision, and most of the observed variance reflects true differences across areas.

Risk adjustment for age and sex appears to affect the areas with the highest and lowest raw rates. Areas with high rates may wish to examine the clinical characteristics of their patients to check for a more complex case mix. Patient age, clinical measures such as heart function, and other management issues may affect admission rates.

As the causes for admissions may include poor quality care, lack of patient compliance, or problems accessing care, areas may wish to review CHF patient records to identify precipitating causes and potential targets for intervention.

Limitations on Use

As a PQI, CHF is not a measure of hospital quality, but rather one measure of outpatient and other health care. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting.

Some CHF care takes place in emergency rooms. As such, combining inpatient and emergency room data may give a more accurate picture of this indicator.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Physician management of patients with congestive heart failure differs significantly by physician specialty.\(^\text{62, 63}\) Such differences in community practices may be reflected in differences in CHF admission rates.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Relatively precise estimates of admission rates for CHF can be obtained, although random variation may be important for small hospitals and rural areas. Based on empirical evidence, this indicator is very precise, with a raw area level rate of 521.0 per 100,000 population and a standard deviation of 286.5.


The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is very high, at 93.0%, indicating that the observed differences in age-sex adjusted rates very likely represent true differences across areas.

**Minimum bias:** Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Important determinants of outcomes with CHF include certain demographic variables, such as patient age; clinical measures; management issues; and treatment strategies. Limited evidence exists on the extent to which these factors can explain area differences in CHF admission rates. Empirical results show that area rankings and absolute performance are somewhat affected by age-sex risk adjustment.

**Construct validity:** Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 4.6 times more CHF hospitalizations per capita than high-income ZIP codes. Millman et al. reported that low-income ZIP codes had 6.1 times more CHF hospitalizations per capita than high-income ZIP codes.

Based on empirical results, areas with high rates of CHF also tend to have high rates of admission for other ACSCs.

**Fosters true quality improvement:** Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Outpatient interventions such as the use of protocols for ambulatory management of low-severity patients and improvement of access to outpatient care would most likely decrease inpatient admissions for CHF.

**Prior use:** Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was originally developed by Billings et al. in conjunction with the United Hospital Fund of New York. It was subsequently adopted by the Institute of Medicine and has been widely used in a variety of studies of avoidable hospitalizations.

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5.7 Low Birth Weight Rate (PQI 9)

Infants may be low birth weight because of inadequate interuterine growth or premature birth. Risk factors include sociodemographic and behavioral characteristics, such as low income and tobacco use during pregnancy.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper preventive care may reduce incidence of low birth weight, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Number of low birth weight infants per 100 births.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Number of births with ICD-9-CM diagnosis codes for birth weight less than 2500 grams in any field.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution.</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Neonates (discharges) in Metro Area or county with age at admission of 0 to 28 days, with ICD-9-CM diagnosis code for in-hospital live birth.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 5.8 per 100 eligible births Rating: 11 out of 16 (Bias was not tested because adequate risk adjustment for low birth weight was not available.) (Smoothing recommended)</td>
</tr>
</tbody>
</table>

Summary of Evidence

Low birth weight is a PQI that would be of most interest to comprehensive health care delivery systems. Healthy People 2010 has set a goal of reducing the percentage of low birth weight infants to 0.9%.68

Mothers who give birth to low birth weight infants generally receive less prenatal care than others, and prenatal care persists as a risk factor for low birth weight when adjusting for potential confounds. However, comprehensive care programs in high-risk women have failed to reduce low birth weights. In some studies, specific counseling aimed at reducing a specific risk factor in a specific population may have some impact on reducing low birth weight.

Adequate risk adjustment may require linkage to birth records, which record many of the sociodemographic and behavioral risk factors noted in the literature review (race, age, drug use, stress). Birth records in some States are a rich source of information that could help to identify causes of low birth weight and help to delineate potential areas of intervention.

Where risk adjustment is not possible, results may provide some guidance to case mix in the area if considered in light of measures of socioeconomic status (as determined by insurance status or ZIP code).

Limitations on Use

As a PQI, low birth weight is not a measure of hospital quality, but rather one measure of outpatient and other health care. This indicator could have substantial bias that would require additional risk adjustment from birth records or clinical data.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Risk factors for low birth weight may be addressed with adequate prenatal care and education. Prenatal education and care programs have been established to help reduce low birth weight and other complications in high-risk populations.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Although low birth weight births account for only a small fraction of total births, the large number of births suggest that this indicator should be precisely measurable for most areas. Based on

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empirical evidence, this indicator is precise, with a raw area level rate of 3.9% and a standard deviation of 2.3%. The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is moderate, at 67.1%, indicating that some of the observed differences in age-sex adjusted rates do not represent true differences in area performance.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Socioeconomic measures such as parental education and income have been shown to be negatively associated with rates of low birth weight infants. Demographic factors such as age and race also appear important, and may be correlated with socioeconomic factors. Mothers under 17 years and over 35 years are at a higher risk of having low birth weight infants. One study of all California singleton births in 1992 found that after risk adjustment, having a black mother remained a significant risk factor. Little evidence exists on the extent to which each of these factors contributes to differences in the rate of low birth weight births across geographic areas.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

While specific studies have demonstrated an impact of particular interventions, especially in high-risk populations, evidence on the impact of better prenatal care on low birth weight rates for area populations is less well developed. In one study, the use of prenatal care accounted for less than 15% of the differences between low birth weight in black and white mothers enrolled in an HMO. However, increasing the level of prenatal care was associated with lower rates of low birth weight, particularly in the black patient population.

Low birth weight is inversely related to the other ACSCs and is positively related to perforated appendix rate. Empirical evidence suggests that this indicator at an area level could be potentially biased.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Use of this indicator is unlikely to lead to apparent reductions in the rate of low birth weight births that did not represent true reductions.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

Low birth weight is an indicator in the Health Plan Employer Data and Information Set (HEDIS) measure set for insurance groups and is used by United Health Care and the University Hospital Consortium. This indicator, along with very low birth weight, was previously an HCUP QI.

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5.8 Dehydration Admission Rate (PQI 10)

Dehydration is a serious acute condition that occurs in frail patients and patients with other underlying illnesses following insufficient attention and support for fluid intake. Dehydration can for the most part be treated in an outpatient setting, but it is potentially fatal for elderly, very young children, frail patients, or patients with serious comorbid conditions.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for dehydration, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for dehydration per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis code for hypovolemia (276.5).</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 127.7 per 100,000 population               Rating: 14</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for dehydration is a PQI that would be of most interest to comprehensive health care delivery systems. Admission for dehydration is somewhat common, suggesting that the indicator will be measured with adequate precision, and most of the observed variation is likely to reflect true differences in admission rates.

This indicator is subject to minimal bias. Risk adjustment appears to affect modestly the areas with the highest and lowest rates. Age may be a particularly important factor, and the indicator should be risk-adjusted for age. Areas with high rates of dehydration admissions also tend to have high rates of admission for other ACSCs.

The considerable variations across areas suggest opportunities for quality improvement in care for patients at risk for dehydration. When high rates of dehydration are identified for a particular hospital, additional study may uncover problems in primary or emergency care in the surrounding area. Appropriate interventions can be developed to address those problems.

Limitations on Use

As a PQI, dehydration is not a measure of hospital quality, but rather one of the measures of outpatient and other health care.

Summary of Evidence

This indicator has unclear construct validity, because it has not been validated except as part of a set of indicators. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting. Some dehydration care takes place in emergency rooms. As such, combining inpatient and emergency room data may give a more accurate picture of this indicator.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Dehydration is a potentially fatal condition, and appropriate attention to fluid status can prevent the condition. If left untreated in older adults, serious complications, including death (over 50%), can result.75

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Little evidence exists in the literature on the precision of this indicator. Based on empirical evidence, this indicator is precise, with a raw
area level rate of 139.9 per 100,000 population and a standard deviation of 103.2.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is high, at 88.5%, indicating that the observed differences in age-sex adjusted rates likely represent true differences across areas.

**Minimum bias:** Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

The age structure of the population may affect admission rates for this condition, as the elderly and very young are more susceptible to dehydration. Socioeconomic factors may also affect admission rates. Differences in thresholds for admission of patients with dehydration may contribute to area rate differences. Empirical results show that area rankings are not affected by age-sex risk adjustment.

**Construct validity:** Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 2 times more dehydration hospitalizations per capita than high-income ZIP codes. Household income explained 42% of this variation. In addition, Millman et al. reported that low-income ZIP codes had 2 times more dehydration hospitalizations per capita than high-income ZIP codes.

Based on empirical results of this study, areas with high rates of dehydration admissions also tend to have high rates of admission for other ACSCs.

**Fosters true quality improvement:** Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Use of this indicator might lead to higher thresholds of admission for patients with dehydration, potentially denying needed care to some patients. Because some dehydration can be managed on an outpatient basis, a shift to outpatient care may occur.

**Prior use:** Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was originally developed by Billings et al. in conjunction with the United Hospital Fund of New York.

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5.9 Bacterial Pneumonia Admission Rate (PQI 11)

Bacterial pneumonia is a relatively common acute condition, treatable for the most part with antibiotics. If left untreated in susceptible individuals—such as the elderly—pneumonia can lead to death.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for bacterial pneumonia in non-susceptible individuals, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for bacterial pneumonia per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis code for bacterial pneumonia.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients with sickle cell anemia or HB-S disease, patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 420.7 per 100,000 population</td>
</tr>
<tr>
<td></td>
<td>Rating: 17</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for bacterial pneumonia is a PQI that would be of most interest to comprehensive health care delivery systems. High admission rates may reflect a large number of inappropriate admissions or low-quality treatment with antibiotics. Admission for pneumonia is relatively common, suggesting that the indicator will be measured with good precision, and most of the observed variation reflects true differences in admission rates.

This indicator is subject to some moderate bias, and risk adjustment appears to affect the areas with the highest rates the most. Age may be a particularly important factor, and the indicator should be risk-adjusted for this factor. Areas may wish to examine the outpatient care for pneumonia and pneumococcal vaccination rates to identify potential processes of care that may reduce admission rates. The patient populations served by hospitals that contribute the most to the overall area rate for pneumonia may be a starting point for interventions.

Limitations on Use

As a PQI, admission for bacterial pneumonia is not a measure of hospital quality, but rather one measure of outpatient and other health care.

This indicator has unclear construct validity, because it has not been validated except as part of a set of indicators. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting. Because some pneumonia care takes place in an emergency room setting, combining inpatient and emergency room data may give a more accurate picture of this indicator.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Vaccination for pneumococcal pneumonia in the elderly and early management of bacterial respiratory infections on an ambulatory basis may reduce admissions with pneumonia. A vaccine developed for the elderly has been shown to be 45% effective in preventing hospitalizations during peak seasons.¹⁸

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Little evidence exists in the literature on the precision or variation in pneumonia admission

rates. Based on empirical evidence, this indicator is precise, with a raw area level rate of 395.6 per 100,000 population and a standard deviation of 208.5.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is very high, at 92.9%, indicating that the observed differences in age-sex adjusted rates likely represent true differences across areas. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

A review of the literature suggests that comorbidities or other risk factors that may vary systematically by area do not significantly affect the incidence of hospitalization for pneumonia. Differences in thresholds for admission of patients with bacterial pneumonia may contribute to area rate differences. Empirical results show that area rankings and absolute performance are somewhat affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 5.4 times more pneumonia admissions per capita than high-income ZIP codes. household income explained 53% of this variation. In addition, Millman et al. reported that low-income ZIP codes had 5.4 times more pneumonia hospitalizations per capita than high-income ZIP codes.

Based on empirical results, areas with high rates of bacterial pneumonia admissions also tend to have high rates of admissions for other ACSCs.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Use of this indicator might lead to higher thresholds of admission for pneumonia patients. Because pneumonia can be managed on an outpatient basis, a shift to outpatient care may occur, which might be inappropriate for more severely ill patients.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was included in Weissman’s set of avoidable hospitalizations.

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81 Weissman JS, Gatsonis C, Epstein AM. Rates of avoidable hospitalization by insurance status in Massachusetts and Maryland JAMA 1992;268(17)2388-94.
5.10 Urinary Tract Infection Admission Rate (PQI 12)

Urinary tract infection is a common acute condition that can, for the most part, be treated with antibiotics in an outpatient setting. However, this condition can progress to more clinically significant infections, such as pyelonephritis, in vulnerable individuals with inadequate treatment.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for urinary infection, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for urinary tract infection per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis code for urinary tract infection.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients with any diagnosis code of kidney or urinary tract disorder or immunocompromised state, transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 170.2 per 100,000 population                      Rating: 11</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for urinary tract infection is a PQI that would be of most interest to comprehensive health care delivery systems. Admission for urinary tract infection is uncommon, but the observed variation is likely to reflect true differences across areas.

Risk adjustment appears to affect the areas with the highest rates the most, and using this indicator without risk adjustment may result in the misidentification of some areas as outliers. This indicator is subject to some moderate bias and should be adjusted for age and sex. The patient populations served by hospitals that contribute the most to the overall area rate for urinary tract infection may be a starting point for interventions.

Limitations on Use

As a PQI, admission for urinary tract infection is not a measure of hospital quality, but rather one measure of outpatient and other health care. This indicator has unclear construct validity, because it has not been validated except as part of a set of indicators. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting. Some urinary tract infection care takes place in emergency rooms. As such, combining inpatient and emergency room data may give a more accurate picture of this indicator.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Uncomplicated urinary tract infections can be treated with antibiotics in the ambulatory setting; however, inappropriate treatment can lead to more serious complications. Admission for urinary tract infection among children, which is rare, is associated with physiological abnormalities.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Little evidence exists in the literature on the precision and variation associated with this indicator. Based on empirical evidence, this indicator is precise, with a raw area level rate of 145.1 per 100,000 population and a standard deviation of 89.5. The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is high, at 84.9%, indicating that the observed differences in age-sex adjusted rates likely
represent true differences across areas. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Differences in thresholds for admission of patients with urinary tract infection may contribute to area rate differences. Empirical results show that area rankings and absolute performance are somewhat affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 2.2 times more urinary tract infection admissions than high-income ZIP codes.\textsuperscript{82} Household income explained 28% of this variation. In addition, Millman et al.\textsuperscript{83} reported that low-income ZIP codes had 2.8 times more urinary tract infection hospitalizations per capita than high-income ZIP codes.

Based on empirical results, areas with high admission rates for urinary tract infections also tend to have high admission rates for other ACSCs.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Use of this indicator might lead to higher thresholds of admission for patients with urinary tract infections.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was originally developed by Billings et al. in conjunction with the United Hospital Fund of New York. It is included in Weissman’s set of avoidable hospitalizations.\textsuperscript{84}


\textsuperscript{84}Weissman JS, Gatsonis C, Epstein AM. Rates of avoidable hospitalization by insurance status in Massachusetts and Maryland. JAMA 1992;268(17):2388-94.
5.11 Angina without Procedure Admission Rate (PQI 13)

Both stable and unstable angina are symptoms of potential coronary artery disease. Effective management of coronary disease reduces the occurrence of major cardiac events such as heart attacks, and may also reduce admission rates for angina.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce admissions for angina (without procedures), and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for angina (without procedures) per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for angina. Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude discharges with a procedure code for cardiac procedure, patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
</tbody>
</table>
| Empirical Results and Rating | Rate (2003): 45.9 per 100,000 population  
Rating: 19                                                                 |

**Summary of Evidence**

Hospital admission for angina is a PQI that would be of most interest to comprehensive health care delivery systems. Admission for angina is relatively common, suggesting that the indicator will be measured with good precision. The observed variation likely reflects true differences in area performance.

Age-sex adjustment has a moderate impact. Other risk factors for consideration include smoking, hyperlipidemia, hypertension, diabetes, and socioeconomic status. The patient populations served by hospitals that contribute the most to the overall area rate for angina may be a starting point for interventions.

**Limitations on Use**

As a PQI, angina without procedure is not a measure of hospital quality, but rather one measure of outpatient and other health care. This indicator has unclear construct validity, because it has not been validated except as part of a set of indicators. Providers may reduce admission rates without actually improving quality of care by shifting care to an outpatient setting. Some angina care takes place in emergency rooms. Combining inpatient and emergency room data may give a more accurate picture.

**Details**

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

Stable angina can be managed in an outpatient setting using drugs such as aspirin and beta blockers, as well as advice to change diet and exercise habits. Effective treatments for coronary artery disease reduce admissions for serious complications of ischemic heart disease, including unstable angina.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Reasonably precise estimates of area angina rates should be feasible, as one study shows that unstable angina accounts for 16.3% of total admissions for ACSCs. Based on empirical evidence, this indicator is adequately precise.

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with a raw area level rate of 166.0 per 100,000 population and a standard deviation of 135.7.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is very high, at 91.6%, indicating that the observed differences in age-sex adjusted rates likely represent true differences across areas. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

No evidence exists in the literature on the potential bias of this indicator. The incidence of angina is related to age structure and risk factors (smoking, hyperlipidemia, hypertension, diabetes) in a population. Elderly age (over 70), diabetes, and hypertension have also been associated with being at higher risk for angina. 87

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 2.3 times more angina hospitalizations than high-income ZIP codes. 88 Household income explained 13% of this variation. In addition, Millman et al. 89 reported that low-income ZIP codes had 2.7 times more angina hospitalizations per capita than high-income ZIP codes.

Based on empirical study, areas with high rates of angina admissions tend to have higher rates of other ACSC admissions.


5.12 Uncontrolled Diabetes Admission Rate (PQI 14)

Uncontrolled diabetes should be used in conjunction with short-term complications of diabetes, which include diabetic ketoacidosis, hyperosmolarity, and coma.*

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment and adherence to care may reduce the incidence of uncontrolled diabetes, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for uncontrolled diabetes per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for uncontrolled diabetes, without mention of a short-term or long-term complication.</td>
</tr>
<tr>
<td></td>
<td>Age 18 years and older.</td>
</tr>
<tr>
<td></td>
<td>Exclude patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 21.0 per 100,000 population Rating: 14</td>
</tr>
</tbody>
</table>

* This indicator is designed to be combined with “Short Term Diabetes Complication Admission Rate” to create the Healthy People 2010 indicator. To do so, users may simply add the rates of the two indicators together.

Summary of Evidence

Hospital admission for uncontrolled diabetes is a PQI that would be of most interest to comprehensive health care delivery systems. Healthy People 2010 has established a goal to reduce the hospitalization rate for uncontrolled diabetes in persons 18-64 years of age from 7.2 per 10,000 population to 5.4 per 10,000 population. Combining this indicator with the short-term diabetes indicator will result in the Healthy People 2010 measure, except that this QI excludes transfers from another institution to reduce double counting of cases. As a result the rate for the AHRQ QI may be minimally lower than the Healthy People 2010 indicator.

This indicator is moderately precise. The observed differences across areas likely reflect true differences in area performance. Age-sex adjustment slightly changes area rankings.

Limitations on Use

As a PQI, uncontrolled diabetes is not a measure of hospital quality, but rather one measure of outpatient and other health care.

Rates of diabetes may vary systematically by area, creating bias for this indicator.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

High-quality outpatient management of diabetic patients has been shown to lead to reductions in almost all types of serious avoidable hospitalizations. However, tight control may be associated with more episodes of hypoglycemia that lead to more admissions.

Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?

Based on empirical evidence, this indicator is moderately precise, with a raw area level rate of 34.7 per 100,000 population and a standard deviation of 28.1.

The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is high, at 72.6%, indicating that the observed differences in age-related variation are largely due to true differences in performance.
sex adjusted rates likely represent true differences in area performance. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Minorities have higher rates of diabetes, and higher hospitalization rates may result in areas with higher minority concentrations. Empirical results show that area rankings in the highest and lowest deciles are slightly affected by age-sex adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Based on empirical results, areas with high rates of uncontrolled diabetes also tend to have high rates of admission for other ACSCs.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

Because diabetic emergencies are potentially life-threatening, hospitals are unlikely to fail to admit patients requiring hospitalization.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This measure corresponds closely with the measure of short-term diabetes that was developed by Billings et al. and described in this document. The key exception is the ICD-9-CM codes 25002 and 25003, which are the only codes included for uncontrolled diabetes.

5.13 Adult Asthma Admission Rate (PQI 15)

Asthma is one of the most common reasons for hospital admission and emergency room care. Most cases of asthma can be managed with proper ongoing therapy on an outpatient basis. Most published studies combine admission rates for children and adults; therefore, areas may wish to examine this indicator together with pediatric asthma.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper outpatient treatment may reduce the incidence or exacerbation of asthma requiring hospitalization, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for adult asthma per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM principal diagnosis codes for asthma.</td>
</tr>
<tr>
<td>Age 18 years and older</td>
<td></td>
</tr>
<tr>
<td>Exclude patients with any diagnosis code of cystic fibrosis and anomalies of the respiratory system, transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
<td></td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 125.1 per 100,000 population</td>
</tr>
<tr>
<td>Rating: 16</td>
<td></td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admission for asthma is a PQI that would be of most interest to comprehensive health care delivery systems.

Environmental factors such as air pollution, occupational exposure to irritants, or other exposure to allergens have been shown to increase hospitalization rates or exacerbate asthma symptoms. While race has been shown to be associated with differences in admission rates, it is unclear whether this is due to differences in severity of disease or inadequate access to care. Adjustment for race is recommended.

Admission rates have been associated with lower socioeconomic status. Areas may wish to identify hospitals that contribute the most to the overall area rate for this indicator. The patient populations served by these hospitals may be a starting point for interventions.

Limitations on Use

As a PQI, adult asthma is not a measure of hospital quality, but rather one measure of outpatient and other health care. Providers may reduce admission rates without actually improving quality by shifting care to an outpatient setting.

Admission rates that are drastically below or above the average or recommended rates should be further examined.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

According to the National Asthma Education Program, asthma is a readily treatable chronic disease that can be managed effectively in the outpatient setting. Observational studies offer some evidence that inhaled steroids may decrease risk of admission by up to 50%. Admission rates that are drastically below or above the average or recommended rates should be further examined.

Asthma is a common cause of admission for adults, and as such this measure is likely to have adequate precision. Based on empirical evidence, this indicator is adequately precise, with a raw area level rate of 107.9 per 100,000 population and a standard deviation of 81.7. The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is high, at 83.6%, indicating that the observed differences in age-sex adjusted rates likely represent true differences across areas.

Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?

Numerous environmental risk factors for asthma have been identified, including allergens, tobacco smoke, and outdoor air pollution. Race represents one of the most complex potentially biasing factors for this indicator. Black patients have consistently been shown to have higher asthma admission rates, even when stratifying for income and age. Adjustment for race is recommended. Empirical results show that area rankings are not affected by age-sex risk adjustment.

Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?

Billings et al. found that low-income ZIP codes in New York City had 6.4 times more asthma hospitalizations than high-income ZIP codes. Household income explained 70% of this variation. In addition, Millman et al. reported that low-income ZIP codes had 5.8 times more asthma hospitalizations per capita than high-income ZIP codes.

Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?

There is little evidence to suggest that asthmatics are being inappropriately denied admission to the hospital. However, because some asthma can be managed on an outpatient basis, a shift to outpatient care may occur.

Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?

This indicator was originally developed by Billings et al. in conjunction with the United Hospital Fund of New York, and is included in Weissman’s set of avoidable hospitalizations.

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5.14 Rate of Lower-extremity Amputation among Patients with Diabetes (PQI 16)

Diabetes is a major risk factor for lower-extremity amputation, which can be caused by infection, neuropathy, and microvascular disease.

<table>
<thead>
<tr>
<th>Relationship to Quality</th>
<th>Proper and continued treatment and glucose control may reduce the incidence of lower-extremity amputation, and lower rates represent better quality care.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Benchmark</td>
<td>State, regional, or peer group average.</td>
</tr>
<tr>
<td>Definition</td>
<td>Admissions for lower-extremity amputation in patients with diabetes per 100,000 population.</td>
</tr>
<tr>
<td>Outcome of Interest</td>
<td>Discharges with ICD-9-CM procedure codes for lower-extremity amputation in any field and diagnosis code for diabetes in any field.</td>
</tr>
<tr>
<td>Age 18 years and older</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Exclude discharges with trauma, patients transferring from another institution, or MDC 14 (pregnancy, childbirth, and puerperium).</td>
</tr>
<tr>
<td>Population at Risk</td>
<td>Population in Metro Area or county, age 18 years and older.</td>
</tr>
<tr>
<td>Empirical Results and Rating</td>
<td>Rate (2003): 36.6 per 100,000 population Rating:10 (Smoothing recommended)</td>
</tr>
</tbody>
</table>

Summary of Evidence

Hospital admissions for lower-extremity amputation among patients with diabetes is a PQI that would be of most interest to comprehensive health care delivery systems.

Lower-extremity amputation (LEA) affects up to 15% of all patients with diabetes in their lifetimes. A combination of factors may lead to this high rate of amputation, including minor trauma to the feet, which is caused by loss of sensation and may lead to gangrene. Proper long-term glucose control, diabetes education, and foot care are some of the interventions that can reduce the incidence of infection, neuropathy, and microvascular diseases. Healthy People 2010 has set a goal of reducing the number of LEAs to 1.8 per 1,000 persons with diabetes.

Studies have shown that LEA varies by age and sex, and age-sex risk adjustment affects moderately the relative performance of areas. Race may bias the indicator, since the rates of diabetes and poor glycemic control are higher among Native Americans and Hispanic Americans. However, results must be interpreted with care when adjusting for race, because poor quality care may also vary systematically with racial composition.

Limitations on Use

As a PQI, lower-extremity amputations among patients with diabetes is not a measure of hospital quality, but rather one measure of outpatient and other health care. PQIs are correlated with each other and may be used in conjunction as an overall examination of outpatient care.

Details

Face validity: Does the indicator capture an aspect of quality that is widely regarded as important and subject to provider or public health system control?

In the United States, diabetes is the leading cause of nontraumatic amputations (approximately 57,000 per year). Possible interventions include foot clinics, wearing proper

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footwear, and proper care of feet and foot ulcers.\textsuperscript{103}

**Precision: Is there a substantial amount of provider or community level variation that is not attributable to random variation?**

Based on empirical evidence, this indicator is moderately precise, with a raw area level rate of 30.5 per 100,000 population and a substantial standard deviation of 42.7. The signal ratio (i.e., the proportion of the total variation across areas that is truly related to systematic differences in area performance rather than random variation) is moderate, at 68.5%, indicating that some of the observed differences in age-sex adjusted rates likely do not represent true differences in area performance. Using multivariate signal extraction techniques appears to have little additional impact on estimating true differences across areas.

**Minimum bias: Is there either little effect on the indicator of variations in patient disease severity and comorbidities, or is it possible to apply risk adjustment and statistical methods to remove most or all bias?**

Several sociodemographic variables are associated with the risk of lower-extremity amputation, including age, duration of diabetes, and sex.\textsuperscript{104,105} Empirical results show that age-sex adjustment affects the relative performance of areas.

**Construct validity: Does the indicator perform well in identifying true (or actual) quality-of-care problems?**

Several studies of intervention programs have noted a decrease in amputation risk. One recent study noted a 1-year post-intervention decrease of 79% in amputations in a low-income African American population. Interventions included foot care education, assistance in finding properly fitting footwear, and prescription footwear.\textsuperscript{106} One observational study found that patients who receive no outpatient diabetes education have a three-fold higher risk of amputation than those receiving care.\textsuperscript{107}

**Fosters true quality improvement: Is the indicator insulated from perverse incentives for providers to improve their reported performance by avoiding difficult or complex cases, or by other responses that do not improve quality of care?**

Given the severity of conditions requiring lower-extremity amputation, hospitals are unlikely to fail to admit patients requiring hospitalization.

**Prior use: Has the measure been used effectively in practice? Does it have potential for working well with other indicators?**

This indicator is not widely used; however, it is included in the DEMPAQ measure set for outpatient care.

\textsuperscript{103}Pecoraro et al. 1990.
\textsuperscript{104}Mayfield et al. 1998.
6.0 Using Different Types of QI Rates

When should you use the observed, expected, risk adjusted, and/or smoothed rates generated by the AHRQ QI software? Here are some guidelines.

If the user’s primary interest is to identify cases for further follow-up and quality improvement, then the observed rate would help to identify them. The observed rate is the raw rate generated by the QI software from the data the user provided. Areas for improvement can be identified by the magnitude of the observed rate compared to available benchmarks and/or by the number of patients impacted.

Additional breakdowns by the default patient characteristics used in stratified rates (e.g., age, gender, or payer) can further identify the target population. Target populations can also be identified by user-defined patient characteristics supplemented to the case/discharge level flags. Trend data can be used to measure change in the rate over time.

Another approach to identify areas to focus on is to compare the observed and expected rates. The expected rate is the rate the provider would have if it performed the same as the reference population given the provider’s actual case-mix (e.g., age, gender, DRG, and comorbidity categories).

If the observed rate is higher than the expected rate (i.e., the ratio of observed/expected is greater than 1.0, or observed minus expected is positive), then the implication is that the provider performed worse than the reference population for that particular indicator. Users may want to focus on these indicators for quality improvement.

If the observed rate is lower than the expected rate (i.e., the ratio of observed/expected is less than 1.0, or observed minus expected is negative), then the implication is that the provider performed better than the reference population. Users may want to focus on these indicators for identifying best practices.

Users can also compare the expected rate to the population rate reported in the detailed evidence section of the IQI, PQI, or PSI Guide to determine how their case-mix compares to the reference population. If the population rate is higher than the expected rate, then the provider’s case-mix is less severe than the reference population. If the population rate is lower than the expected rate, then the provider’s case-mix is more severe than the reference population.

We use this difference between the population rate and the expected rate to “adjust” the observed rate to account for the difference between the case-mix of the reference population and the provider’s case-mix. This is the provider’s risk-adjusted rate.

If the provider has a less severe case-mix, then the adjustment is positive (population rate > expected rate) and the risk-adjusted rate is higher than the observed rate. If the provider has a more severe case-mix, then the adjustment is negative (population rate < expected rate) and the risk-adjusted rate is lower than the observed rate. The risk-adjusted rate is the rate the provider would have if it had the same case-mix as the reference population given the provider’s actual performance.

Finally, users can compare the risk-adjusted rate to the smoothed or “reliability-adjusted” rate to determine whether this difference between the risk-adjusted rate and reference population rate is likely to remain in the next measurement period. Smoothed rates are weighted averages of the population rate and the risk-adjusted rate, where the weight reflects the reliability of the provider’s risk-adjusted rate.

A ratio of (smoothed rate - population rate) / (risk-adjusted rate - population rate) greater than 0.80 suggests that the difference is likely to persist (whether the difference is positive or negative). A ratio less than 0.80 suggests that the difference may be due in part to random differences in patient characteristics (patient characteristics that are not observed and controlled for in the risk-adjustment model). In general, users may want to focus on areas where the differences are more likely to persist.

PQI Guide 48 Version 3.0a (February 20, 2006)
7.0 References


Appendix A: Links

The following links may be helpful to users of the AHRQ Prevention Quality Indicators.

Prevention Quality Indicators Version 3.0a Documents and Software

Available at [http://www.qualityindicators.ahrq.gov/pqi_download.htm](http://www.qualityindicators.ahrq.gov/pqi_download.htm)

<table>
<thead>
<tr>
<th>Title</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Guide to Prevention Quality Indicators</td>
<td>Describes how the PQIs were developed and provides detailed evidence for each indicator.</td>
</tr>
<tr>
<td>Prevention Quality Indicators Technical Specifications</td>
<td>Provides detailed definitions of each PQI, including all ICD-9-CM and DRG codes that are included in or excluded from the numerator and denominator. Note that exclusions from the denominator are automatically applied to the numerator.</td>
</tr>
<tr>
<td>PQI Covariates used in Risk Adjustment</td>
<td>Tables for each PQI provide the stratification and coefficients used to calculate the risk-adjusted rate for each strata.</td>
</tr>
<tr>
<td>SAS® PQI Software Documentation</td>
<td>This software documentation provides detailed instructions on how to use the SAS® version of the PQI software including data preparation, calculation of the PQI rates, and interpretation of output.</td>
</tr>
<tr>
<td>SPSS® PQI Software Documentation</td>
<td>This software documentation provides detailed instructions on how to use the SPSS® version of the PQI software including data preparation, calculation of the PQI rates, and interpretation of output.</td>
</tr>
<tr>
<td>Change Log to PQI Documents and Software</td>
<td>The Change Log document provides a cumulative summary of all changes to the PQI software, software documentation, and other documents made since the release of version 2.1 of the software in March 2003. Changes to indicator specifications that were not a result of new ICD-9-CM and DRG codes, are also described in the Change Log.</td>
</tr>
<tr>
<td>Fiscal year 2006 Coding Changes</td>
<td>This document summarizes the changes to the indicator definitions resulting from FY 2006 changes to ICD-9-CM coding and DRG changes. These changes will only affect data from FY 2006 (October 1, 2005) or later.</td>
</tr>
<tr>
<td>SAS® PQI Software</td>
<td>Requires the SAS® statistical program distributed by the SAS Institute, Inc. The company may be contacted directly regarding the licensing of its products: <a href="http://www.sas.com">http://www.sas.com</a></td>
</tr>
<tr>
<td>SPSS® PQI Software</td>
<td>Requires the SPSS® statistical program distributed by SPSS, Inc. The company may be contacted directly regarding the licensing of its products: <a href="http://www.spss.com">http://www.spss.com</a></td>
</tr>
</tbody>
</table>
AHRQ QI Windows Application

The AHRQ QI Windows Application calculates rates for all of the AHRQ Quality Indicators modules and does not require either SAS® or SPSS®. It is available at:

http://www.qualityindicators.ahrq.gov/winqi_download.htm

Additional Documents

The following documents are available within the "Documentation" section of the AHRQ QI Downloads Web page:

http://www.qualityindicators.ahrq.gov/downloads.htm

- Refinement of the HCUP Quality Indicators (Technical Review), May 2001
- Refinement of the HCUP Quality Indicators (Summary), May 2001
- Measures of Patient Safety Based on Hospital Administrative Data - The Patient Safety Indicators, August 2002
- Measures of Patient Safety Based on Hospital Administrative Data - The Patient Safety Indicators (Summary), August 2002

In addition, these documents may be accessed at the AHRQ QI Documentation Web page:

http://www.qualityindicators.ahrq.gov/documentation.htm

- Guidance for Using the AHRQ Quality Indicators for Hospital-level Public Reporting or Payment, August 2004
- AHRQ Summary Statement on Comparative Hospital Public Reporting, December 2005
- Appendix A: Current Uses of AHRQ Quality Indicators and Considerations for Hospital-level
- Comparison of Recommended Evaluation Criteria in Five Existing National Frameworks

The following documents can be viewed or downloaded from the page:

http://www.qualityindicators.ahrq.gov/newsletter.htm

- 2006 Area Level Indicator Changes
- Considerations in Public Reporting for the AHRQ QIs
- June 2005 Newsletter - Contains the article, "Using Different Types of QI Rates"

Other Tools and Information

PQI rates can be calculated using the modified Federal Information Processing Standards (FIPS) State/county code. A list of codes is available at:

http://www.census.gov/popest/geographic/codes02.pdf

AHRQ provides a free, on-line query system based on HCUP data that provides access to health statistics and information on hospital stays at the national, regional, and State level. It is available at:

http://hcup.ahrq.gov/HCUPnet.asp

The CDC National Diabetes Surveillance System provides state level estimates of diabetes prevalence by age.