

Medicare Data for the Geographic Variation Public Use File: A Methodological Overview December 2013 Update

Introduction

Federal policymakers and health researchers have long recognized that the amount and quality of the health care services that Medicare beneficiaries receive vary substantially across different regions of the United States. Much of that variation does not appear to be caused by differences in beneficiaries' health, and one widely-publicized estimate asserted that as much as 30 percent of Medicare expenditures may be unnecessary.¹

The Office of Information Products and Data Analytics within the Centers for Medicare & Medicaid Services (CMS) has developed a public use file, the Geographic Variation Public Use File (GV PUF), to support further analysis of this important issue. This public use file is based primarily on information from CMS's Chronic Conditions Data Warehouse (CCW), which contains 100 percent of Medicare claims for beneficiaries who are enrolled in the fee-for-service (FFS) program as well as enrollment and eligibility data. The GV PUF covers calendar years 2008-2012 and has information on the demographics, spending, service utilization, and prevalence of certain chronic conditions for Medicare beneficiaries in different parts of the country. We also incorporated a variety of quality indicators that can be used to analyze relationships between Medicare utilization and quality of care.

The December 2013 update to the GV PUF adds data for 2012 and incorporates several minor revisions to the CMS methodology. This update supersedes the data that we provided in May 2013.

This overview is divided into the following eight sections:

1. Key data sources
2. Study population
3. Geographic variables
4. Disease variables
5. Standardization and risk adjustment of spending
6. Utilization measures
7. Quality measures
8. Changes from the May 2013 dataset to the December 2013 update

1. Key data sources

The primary data source for these data is CMS's Chronic Conditions Data Warehouse (CCW). The CCW contains 100 percent of Medicare claims for beneficiaries who are enrolled in the fee-for-service (FFS) program as well as enrollment and eligibility data. The CCW was designed as

¹ John Wennberg et al. *Tracking the Care of Patients with Severe Chronic Illness – The Dartmouth Atlas of Health Care 2008*, The Dartmouth Institute for Health Policy and Clinical Practice.

a database to support research on chronically ill beneficiaries, so it also contains other valuable features, such as a unique identifier for each beneficiary that makes it possible to track spending for individual beneficiaries over time and flags that indicate if a beneficiary has one or more of 27 specific chronic conditions.

The detailed nature of the CCW claims data makes it possible to analyze differences in cost and/or utilization for specific settings of care or types of services. Some of the settings include inpatient hospital, outpatient hospital, multiple post-acute care settings (long-term care hospital, inpatient rehabilitation facility, skilled nursing facility, and home health agency), hospice, physicians, laboratories, and suppliers of durable medical equipment. The data in this request is for calendar years 2008-2012.

Physician services are defined using the Berenson-Eggers Type of Service (BETOS) classification scheme, which groups services into six major categories: physician evaluation and management, physician procedures, imaging, laboratory tests, durable medical equipment, and other. The total number of distinct BETOS codes is much larger – about 120 – when you count the numerous sub-groupings within those major categories.

We also incorporated several quality measures into the data. Those measures were derived from two publicly available sets of quality measures:

- Hospital Compare (HC), which was developed by CMS and uses data from hospitals and Medicare claims to create measures on inpatient processes, readmissions, and mortality.
- Prevention Quality Indicators (PQIs), which is software developed by the Agency for Healthcare Research and Quality (AHRQ) that uses administrative data to measure hospital admission rates for ambulatory care sensitive conditions.

Both sets of measures are well-known to health care researchers and have been endorsed by the National Quality Forum.

In addition to the quality measures described above, we also calculated the number of times that Medicare beneficiaries visited hospital emergency departments and all-cause hospital readmission rates.

2. Study population

Since the primary goal of the GV PUF is to make it easier to analyze differences in health care utilization and spending for Medicare beneficiaries living in different parts of the United States, we created analytic files that exclude certain categories of Medicare beneficiaries to make those comparisons as meaningful as possible.

Table 1 shows the number and percent of beneficiaries excluded, by year. We applied the same exclusions to each year of the data. Note that whether individual beneficiaries were part of the study population could vary from year to year, depending on whether and when one of the exclusions described below applied to them.

First, we excluded beneficiaries who were enrolled at any point during the year in a Medicare Advantage (MA) plan. (There were 14.7 million beneficiaries in MA plans in 2012, about 28 percent of the overall total.)

Second, we excluded beneficiaries who were enrolled at any point in the year in Part A only or Part B only (roughly 4.7 million in 2012, about 9 percent of the overall total). Since those beneficiaries are enrolled in only one part of Medicare, their per-capita spending cannot be compared directly to spending for beneficiaries that are enrolled in both Part A and Part B.

Although we report data for beneficiaries of all ages, we also report data separately for two age groups: beneficiaries who were under the age of 65 and received Medicare because they were either disabled or had end-stage renal disease (6.2 million in 2012) and beneficiaries age 65 and older (nearly 27.9 million in 2012). We report data separately by age group because beneficiaries under 65 differ in numerous respects from the over-65 population and could have different health service needs that are difficult to adjust for across geographic regions.

We would like to note that our analytic files do include beneficiaries who died during the calendar year (about 4 percent of the study population) as long as they were not excluded for one of the reasons outlined above.

Table 2 provides some basic demographic information about the beneficiaries.

3. Geographic variables

We used hospital referral regions (HRRs), as well as states and counties, as the geographic units of analysis. HRRs were developed by the Dartmouth Atlas of Health Care to delineate regional health care markets in the United States. See Appendix 1 for a complete list of HRRs.

The Dartmouth Atlas constructed HRRs by grouping ZIP codes together based on the referral patterns for tertiary care for Medicare beneficiaries. HRRs also had to have a minimum overall population of 120,000, and the residents of each HRR had to receive at least 65 percent of their hospitalizations within the HRR. There are 306 HRRs in the United States, and their boundaries often cross state lines. For example, the HRR for Memphis, Tennessee, includes parts of southeastern Missouri, eastern Arkansas, and northern Mississippi.

We assigned Medicare spending to HRRs and states based on where beneficiaries live, rather than where they received care. Although HRRs are smaller than states, they are large enough to encompass most of the care received by beneficiaries, even if they obtain care in multiple localities or counties. Our data show that roughly 81 percent of Medicare expenditures in 2012 occurred in the same HRR where the beneficiary lived. Furthermore, HRRs generally have populations that are large enough to generate stable averages for comparisons of cost and utilization, even for narrowly defined combinations of conditions and services.

Table 1: Study Population in the GV PUF

	2008		2009		2010		2011		2012	
	<u>Number</u>	<u>Percent</u>								
Total Medicare beneficiaries	47,850,425	100.0%	48,922,869	100.0%	50,088,947	100.0%	51,717,260	100.0%	53,551,614	100.0%
Beneficiaries excluded:										
Any enrollment in MA	11,010,037	23.0%	12,061,219	24.7%	12,668,675	25.3%	13,423,854	26.0%	14,752,372	27.5%
Part A only or B only	4,217,045	8.8%	4,313,134	8.8%	4,378,633	8.7%	4,555,881	8.9%	4,672,937	8.7%
Total excluded beneficiaries	15,227,082	31.8%	16,374,353	33.5%	17,047,308	34.0%	17,979,735	34.8%	19,425,309	36.3%
Study population	32,623,343	68.2%	32,548,516	66.5%	33,041,639	66.0%	33,737,525	65.2%	34,126,305	63.7%
Beneficiaries in study population that died in the year	1,509,072	4.6%	1,452,296	4.5%	1,473,223	4.5%	1,492,527	4.4%	1,484,460	4.3%

Note: Percentages may not sum to totals because of rounding.

Table 2: Demographics of Beneficiaries in the GV PUF

	2008		2009		2010		2011		2012	
	<u>Number</u>	<u>Percent</u>								
Total Medicare beneficiaries	32,623,343	100.0%	32,548,516	100.0%	33,041,639	100.0%	33,737,525	100.0%	34,126,305	100.0%
By age:										
<40	901,182	2.8%	918,745	2.8%	958,193	2.9%	998,615	3.0%	1,010,600	3.0%
40 to 64	4,704,807	14.4%	4,801,334	14.8%	5,009,681	15.2%	5,190,291	15.4%	5,226,723	15.3%
65 to 74	13,192,455	40.4%	13,202,607	40.6%	13,412,970	40.6%	13,896,888	41.2%	14,367,585	42.1%
75 to 84	9,378,767	28.7%	9,138,225	28.1%	9,080,655	27.5%	9,017,739	26.7%	8,878,555	26.0%
85 to 94	4,029,101	12.4%	4,061,542	12.5%	4,138,498	12.5%	4,182,047	12.4%	4,181,968	12.3%
95+	417,031	1.3%	426,063	1.3%	441,642	1.3%	451,945	1.3%	460,874	1.4%
By gender:										
Female	18,230,611	55.9%	18,145,602	55.7%	18,353,489	55.5%	18,650,223	55.3%	18,799,917	55.1%
Male	14,392,732	44.1%	14,402,914	44.3%	14,688,150	44.5%	15,087,302	44.7%	15,326,388	44.9%
By race/ethnicity:										
White, non-Hispanic	26,657,863	81.7%	26,449,593	81.3%	26,700,882	80.8%	27,136,137	80.4%	27,359,619	80.2%
African-American	3,049,460	9.3%	3,104,574	9.5%	3,214,204	9.7%	3,317,539	9.8%	3,352,937	9.8%
Hispanic	1,834,133	5.6%	1,872,340	5.8%	1,937,957	5.9%	2,006,559	5.9%	2,035,422	6.0%
Asian/Pacific Islander	660,918	2.0%	684,208	2.1%	716,326	2.2%	746,877	2.2%	775,074	2.3%
Other	420,969	1.3%	437,801	1.3%	472,270	1.4%	530,413	1.6%	603,253	1.8%

Note: "Other" includes American Indian/Alaskan Native, other race, and unknown.

4. Disease Variables

The geographic variation in Medicare spending may be due, at least in part, to regional differences in the prevalence of particular diseases (or combinations of diseases). For example, Medicare spending in a particular area could be higher because the beneficiaries who live there are more likely to suffer from an expensive condition such as heart failure.

For this reason, we also include data on prevalence of disease for 19 different chronic conditions that are a standard part of the CCW data. Those conditions are:

- Acute myocardial infarction (heart attack)
- Alzheimer's disease and related disorders
- Arthritis
- Asthma
- Atrial fibrillation
- Chronic kidney disease
- Chronic obstructive pulmonary disease
- Colorectal cancer
- Depression
- Diabetes
- Female breast cancer
- Heart failure
- Hyperlipidemia (high cholesterol)
- Hypertension (high blood pressure)
- Ischemic heart disease
- Osteoporosis
- Lung cancer
- Prostate cancer
- Stroke

The conditions listed above are not mutually exclusive, so they are best suited for measuring the overall prevalence of a particular condition within the Medicare population. At the same time, beneficiaries can (and often do) have more than one condition, and those additional conditions can cause substantial variation in spending and utilization patterns.

5. Standardization and risk adjustment

These data will help users analyze underlying differences in resource use among Medicare beneficiaries in different parts of the country. These differences reflect variation in such factors as physicians' practice patterns and beneficiaries' ability and willingness to obtain care. However, Medicare spending and utilization can vary for reasons that are not attributable to practice patterns or willingness to seek care, and two of those reasons are particularly important. First, Medicare often pays different amounts for the same service in different areas (for example, to reflect variation in local wages or input prices). Second, the health of Medicare beneficiaries also varies geographically, and those differences will clearly affect spending and utilization.

To account for those factors, we modified the data from the CCW in two ways:

- We standardized Medicare's payment amounts to remove geographic differences in payment rates for individual services as a source of variation, and
- We adjusted for differences in beneficiaries' health using the risk-adjustment model that CMS uses to pay MA plans.

Standardization

We standardized payment rates using the same methodology that CMS uses to calculate its Medicare spending per beneficiary (MSPB) metric for the hospital value-based purchasing program. (The only exception is that we standardize the Medicare payment amount, while the methodology used for the MSPB standardizes the allowed amount.) This methodology examines Medicare's various FFS payment systems and identifies the factors that lead to different payment rates for the same service. In general, those factors are adjustments that Medicare makes to account for local wages or input prices, and extra payments that Medicare makes to advance other program goals, such as compensating certain hospitals for the cost of training doctors. We generally then either built up a payment amount using just the base rate for the service and the weight applied under the particular payment system for that service, or worked backward from the actual payment amount to determine what Medicare would have paid without those adjustments.²

The process that we used to calculate standardized payments for each claim under the major FFS payment systems is summarized below. For additional detail, please refer to the Technical Supplement.

Inpatient acute care hospitals paid under the prospective payment system (PPS). We took the operating and capital base rates and multiplied them by the relative weight for each claim's diagnosis-related group. We then added an adjusted outlier payment as well as any new technology payment if they were included on the claim.

Medicare uses the hospital wage index to adjust base rates to reflect local differences in wage levels. For example, the base payment rate in FY 2010 for chronic obstructive pulmonary disease (without any complications or comorbidities) was \$4,056 but the amount that Medicare paid after the wage index was applied ranged from a low of \$3,391 in rural Alabama to a high of \$5,768 in Santa Cruz, California. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Our methodology excluded a number of other payments that hospitals can receive under the PPS: payments for medical education (both direct and indirect), payments to hospitals that serve a disproportionate share of low-income patients, payments for bad debt (deductibles and cost

² If the standardization methodology creates a standardized payment amount for a given claim that exceeds the actual payment amount by \$250,000 or more, we use the actual payment as the standardized amount for that claim. This adjustment impacts less than 300 claims out of the hundreds of millions of claims processed for a given year.

sharing that beneficiaries do not pay), and extra payments to certain rural hospitals such as sole community hospitals and Medicare-dependent hospitals.

Under the inpatient PPS, Medicare uses a per-diem methodology to make reduced payments for certain short-stay transfers and for discharges in certain MS-DRGs where the beneficiary goes on to receive post-acute care. In those instances, we used the reduced payment amount as the starting point for standardization rather than the higher, DRG-based amount.

Payment rates for hospitals in Maryland are set by the state's all-payer rate-setting commission. For claims from those facilities for 2008-2011, we worked backward from actual paid amounts to remove the impact of wages, using the wage index, and of payments for medical education and uncompensated care, using hospital-specific factors that we developed based on data supplied by the state's rate-setting commission. Starting with 2012, we used a different approach and determined the standardized payment using the method for inpatient PPS hospitals.

Critical Access Hospitals (CAHs). For 2008-2011, since CAHs are paid on a cost-plus basis, we determined the standardized payment by adjusting for the lower wages paid in rural areas, which we calculated by dividing the actual amount paid by the local wage index. Starting with 2012, we used a different approach and determined the standardized payment using the method for inpatient PPS hospitals.

Long-term care hospitals (LTCHs). We took the base payment rate for LTCHs and multiplied it by the relative weight for each claim's diagnosis-related group. We then added an adjusted outlier payment if it was included on the claim. As with inpatient PPS hospitals, payments are reduced for certain short stays, and in those cases we used the reduced payment amount as the starting point for standardizing instead of the DRG-based amount.

Medicare uses the hospital wage index to adjust base rates for LTCHs. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Inpatient rehabilitation facilities (IRFs). We took the base payment rate for IRFs and multiplied it by the weight for each claim's case-mix-group. We then added an adjusted outlier payment that the hospital received if the claim was for an unusually high-cost case. As with inpatient PPS hospitals, payments are reduced for certain short stays, and in those cases we used the reduced payment amount as the starting point for standardizing instead of the amount based on the case-mix-group.

Medicare uses the hospital wage index to adjust base rates for IRFs. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Our methodology excluded the additional payments that Medicare makes to certain rural facilities, facilities that treat large numbers of low-income patients, and facilities that are part of teaching hospitals.

Inpatient psychiatric facilities (IPFs). We took the base payment rate for IPFs and followed Medicare's rules for adjusting that rate to account for the patient's age, the weight for their diagnosis-related group, length of stay, and comorbidities (if any). We then added an adjusted outlier payment and a payment for electroconvulsive therapy if the claim included such add-ons.

Medicare uses the hospital wage index to adjust base rates for IPFs. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Our methodology also excludes the additional payments that Medicare makes to certain rural facilities, facilities that are part of teaching hospitals, and facilities in Alaska and Hawaii.

Skilled nursing facilities (SNFs). We took the base daily payment rates for SNFs (there are separate ones for urban and rural facilities; we used the average of the two) and multiplied the nursing and rehabilitation components by the appropriate weight for that claim's resource utilization group. By using the base payment rates, we eliminated the impact of the hospital wage index, which Medicare uses to adjust SNF payment rates. We then multiplied the overall daily rate by the number of days on the claim and the AIDS adjustment, if applicable. We calculated standardized payments for swing beds in CAHs in a manner similar to inpatient services in CAHs (i.e., we adjusted for the effects of local wages).

Home health services. We took the base rate for home health services and multiplied it by the weight for each claim's home health resource group. We then added an adjusted outlier payment that the home health agency received if the claim was for an unusually high-cost case. For short stay claims, we work backward from the actual payment, removing the impact of the wage index.

Medicare uses a version of the hospital wage index to adjust base rates and outlier payments for home health services. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Hospice care. We generally used the base daily and hourly payment rates for hospice care as the standardized rates. By using the base payment rates, we eliminated the impact of the hospice wage index, which Medicare uses to adjust hospice payment rates. For services of physicians or nurse practitioners billed on a hospice claim, we used the actual payment as the standardized amount.

Outpatient hospital services PPS. We calculated standardized payments amounts at the revenue center line level, except for outlier payments, which we determined at the claim level. For revenue center lines that were included in an ambulatory payment classification (APC), we generally used the conversion factor for outpatient services and multiplied it by the weight for the relevant APC and by the number of units. For revenue center lines that were paid using another Part B payment system (such as clinical lab), we used the methodology described below for that payment system. To the extent feasible, we followed payment rules that reduce payment amounts for multiple or interrupted services by 50 percent. For revenue center lines that reflect pass-through services, we did not make any adjustments to the actual paid amounts. Finally, we

added an adjusted outlier payment that the hospital received if the claim was for an unusually high-cost case.

As with inpatient services, payment rates for hospitals in Maryland are set by the state's all-payer rate-setting commission; in those cases, for 2008-2011, we worked backward from actual paid amounts to remove the impact of wages, using the wage index, and of payments for medical education and uncompensated care, using hospital-specific factors that we developed based on data supplied by the state's rate-setting commission. For outpatient services that were provided by CAHs in 2008-2011, we used a methodology similar to the one used for CAH inpatient services during those years. Starting with 2012, for both Maryland hospitals and CAHs, we used a different approach and determined the standardized payment using the method for outpatient PPS hospitals.

Medicare uses the hospital wage index to adjust base rates and outlier payments for outpatient hospital services. Effectively, under standardization we calculated all payment amounts with the wage index set at 1.0 to eliminate those differentials.

Outpatient dialysis facilities. For outpatient dialysis facilities, we determined standardized amounts by working backward from paid amounts to remove the impact of the wage index.

Ambulatory surgical centers (ASCs). We took the conversion factor for ASC services and multiplied it by the relative weight for the ASC service provided and by the number of units provided. By using just the conversion factor and the relative weights, we eliminated the impact of the hospital wage index, which Medicare uses to adjust ASC payment rates. We followed Medicare rules by reducing payment amounts on claims for multiple or interrupted services by 50 percent.

Physician services. Medicare uses three geographic practice cost indices to adjust payment rates for physician services. We eliminated those differentials by simply taking the appropriate facility or non-facility payment amount from the fee schedule. We followed Medicare payment rules such as the reduction for multiple procedures and the reduction when services are provided by non-physician providers (such as physician assistants and nurse practitioners). Standardized payments do not include bonuses received in health professional shortage areas or the discount on payments to non-participating physicians.

Anesthesia services. For anesthesia claims, we used the base time unit, added any additional 15-minute time units, and multiplied the sum by the conversion factor. We followed payment rules with regard to discounting multiple procedures or when services are furnished by a certified registered nurse anesthetist.

Durable medical equipment (DME), prosthetics, and orthotics. Medicare pays for DME, prosthetics, and orthotics using a combination of state-specific fee schedules and a national fee schedule that has minimum and maximum payment amounts. (Since 2011, Medicare has used competitive bidding to pay for certain DME items in some areas.) For DME claims, we used the ceiling amount on the national fee schedule as the payment amount for each claim. For prosthetics and orthotics, we used five-sixths of the ceiling as the payment amount.

Laboratory services. Medicare pays for laboratory services using state-specific fee schedules, but they are subject to a national limitation that applies to most claims. Generally, we used that national limit times the number of units to determine the standardized amount.

Ambulance services. Medicare pays for ambulance services using a fee schedule that pays separately for mileage and for the level of support provided during the trip. We did not make any adjustments to payments for mileage-related codes. For all other codes, we used the average payment amount for each code as the standardized amount.

Other services. We did not adjust payment amounts for drugs covered under Part B (which are paid using national rates) or for parenteral and enteral nutrition claims. For federally-qualified health centers and rural health centers, we worked back from the actual payment amount to determine the standardized amount by removing the impact of wage variation. The determination of standardized payments for Comprehensive Outpatient Rehabilitation Facilities (CORFs) and Outpatient Rehabilitation Facilities (ORFs) follows the methodology for physician services. Determination of standardized payments for Community Mental Health Centers (CMHCs) follows the methodology for hospital outpatient services.

Finally, we reduced all payment amounts to reflect any cost sharing that Medicare beneficiaries paid through a deductible, copayment, or coinsurance. For example, Part A had a deductible in 2012 of \$1,156 for inpatient care and charged copayments on beneficiaries who received more than 60 days of inpatient care, while Part B had a deductible of \$140 and required beneficiaries to pay coinsurance of 20 percent for most services.

Risk adjustment

CMS developed a risk-adjustment model that uses HCCs (hierarchical condition categories) to assign risk scores. Those scores estimate how beneficiaries' FFS spending will compare to the overall average for the entire Medicare population. The average risk score is set at 1.0; beneficiaries with scores greater than that are expected to have above-average spending, and vice versa. Risk scores are based on a beneficiary's age and sex; whether the beneficiary is eligible for Medicaid, first qualified for Medicare on the basis of disability, or lives in an institution (usually a nursing home); and the beneficiary's diagnoses from the previous year.³ The HCC model was designed for risk adjustment on larger populations, such as the enrollees in an MA plan, and generates more accurate results when used to compare groups of beneficiaries rather than individuals.

CMS uses HCCs to determine the diagnosis-related portion of the risk score. For example, the HCC system for 2010 included a total of 189 conditions, with related conditions grouped into 70

³ Other methods of risk adjustment exist. For example, the Dartmouth Atlas has adjusted for risk in some of its research by comparing beneficiaries with the same chronic condition during the last two years of life and by comparing beneficiaries who are admitted to the hospital for the same reason. We decided to use the HCC model because it is generally regarded as the best risk-adjustment model available and is used by CMS for both MA and (in a modified form) Part D payment. However, the HCC model relies in part on diagnoses, so scores may reflect variation in physicians' practice patterns rather than beneficiaries' health status. For example, some areas with high utilization patterns may look riskier because more diagnoses will show up on claims.

disease hierarchies. One hierarchy had three different diseases that affect the liver: end-stage liver disease, cirrhosis, and chronic hepatitis. Each condition had a weight that reflects its marginal contribution to a beneficiary's total expected Medicare costs.

Under the HCC system, CMS calculates the diagnosis-related portion of a beneficiary's risk score by adding up the weights for the most severe diagnosis that the beneficiary has in each disease hierarchy. Continuing the example above, a beneficiary with both cirrhosis (weight = 0.406) and chronic hepatitis (weight = 0.406) would receive credit only for the cirrhosis diagnosis.⁴ The researchers who developed the HCC system adopted this approach after finding that having multiple conditions within a hierarchy did not increase overall patient spending substantially.

We used the risk scores to adjust spending data at the beneficiary level rather than in aggregate. As a result, the aggregate standardized, risk-adjusted spending in a region does not equal the aggregate standardized costs divided by the average HCC risk score. In addition, the HCC model was not designed to risk-adjust spending for individual services and therefore is not applied to service-level spending. The Medicare Payment Advisory Commission has used a similar approach in some of its work.⁵

By standardizing payment amounts and adjusting for differences in beneficiaries' health status, these data provide a more accurate picture of how resource use varies for Medicare beneficiaries across the country.

6. Utilization measures

In addition to standardizing and risk-adjusting spending amounts, we also calculated a series of figures that measure actual utilization for certain major types of Medicare-covered services. We used the claims-level data from the CCW to generate three different types of utilization measures for each geographic region:

- The *number of times* that the beneficiaries in our study population used a particular service, expressed in terms of usage per 1,000 beneficiaries. We calculated these figures across all beneficiaries in our study population, not just the beneficiaries who used that particular service. The metrics that we used to measure utilization varied by the type of service and are described in more detail below.
- The *number of beneficiaries* in our study population who used a particular service
- The *percentage of beneficiaries* in our study population who used a particular service

⁴ The HCC model has two sets of weights: one for beneficiaries living in the community and another for beneficiaries living in an institution. This example uses the weights for a beneficiary living in the community (which happen to be identical for these two conditions).

⁵ For example, see Medicare Payment Advisory Commission, *Measuring Regional Variation in Service Use*, December 2009.

We generated these utilization measures for 15 major service categories. Those categories are listed below, grouped by the units of measurement that we used for each service:

- Number of stays, number of days of care⁶
 - Inpatient hospital care (including inpatient acute care hospitals paid under the PPS, CAHs, and other inpatient hospital care⁷)
 - LTCHs
 - IRFs
 - SNFs
 - Hospice

- Number of episodes, number of visits
 - Home health

- Number of visits
 - Hospital outpatient services
 - Outpatient dialysis facilities
 - Clinics (federally-qualified health centers and rural health centers)

- Number of Events
 - ASCs
 - Physician evaluation and management services
 - Physician procedures
 - Imaging
 - DME
 - Tests (laboratory and non-laboratory)

We also generated figures for the number and percentage of beneficiaries using prescription drugs that are covered under Part B. We did not calculate the number of times that beneficiaries used those drugs because of the difficulty in devising a standard way to measure their utilization.

Finally, we also calculated four metrics on all-cause hospital readmissions⁸ and emergency room (ER) use:

- Total number of all-cause hospital readmissions

- All-cause hospital readmission rate (i.e., the number of readmissions divided by the total number of admissions where the beneficiary was discharged alive)

- Total number of ER visits

- Total number of ER visits per 1,000 beneficiaries

⁶ Our calculations for all hospital-related and SNF services were based only on Medicare-covered days.

⁷ This category includes hospitals such as IPFs and cancer hospitals.

⁸ We used all readmissions that took place within 30 days of the initial discharge.

7. Quality measures

The relationships between the quality, use, and cost of health care are important elements to consider when analyzing the geographic variation in Medicare spending. For example, do areas with above-average spending provide high-quality care, or is there little correlation between the two?

The statistics on hospital readmissions and ER visits discussed above are useful in examining some issues related to the quality of care, such as continuity of care and access to primary care. We have supplemented those metrics by adding dozens of other quality-related measures to support additional analyses. We first selected individual quality measures from two different measure sets:

- Hospital Compare (HC), which was developed by CMS and uses data from hospitals and Medicare claims to measure processes and outcomes for hospital care for heart attack, heart failure, pneumonia, and surgical care. Hospital Compare results are only available in aggregate, so we do not present them by age group. In addition, due to the small cell sizes for many of the measures, we do not present the Hospital Compare measures in the county-level data.
- Prevention Quality Indicators (PQI), which is software developed by AHRQ that uses administrative data to measure hospital admission rates for ambulatory care sensitive conditions. Due to small cell sizes for many of the measures, we do not present the PQIs in the county-level data.

Those measure sets have been endorsed by the National Quality Forum and are well-known to health care researchers and quality improvement organizations. See Appendix 2 for a complete list of the measures that we included in the data set.

Calculation of HRR-level and state-level scores for individual measures. The two data sets contain a total of 51 different measures. We decided not to use twelve of those measures, either because they address issues that are not significant for the Medicare population (such as obstetric care) or because the sample size is too small. We then took the remaining 39 measures, which are usually reported for an individual ZIP code or provider, and aggregated them at the HRR and state level. We did so as follows:

- HC contains both process and outcomes measures. The process measures are based on a sample of each hospital's patients (both Medicare and non-Medicare); we used provider ZIP codes to identify the hospitals in each HRR or state and then calculated a weighted average for the HRR or state using each hospital's patient population for the three primary conditions measured (heart attack, heart failure, and pneumonia) as its weight.

The outcomes measures are based on each hospital's entire Medicare patient population. Those measures have underlying numerators and denominators. For example, the 30-day death rate for heart attack patients has the number of heart attack patients that died as the

numerator and the total number of heart attack patients as the denominator. We added the numerators for all hospitals in a given HRR or state and divided that figure by the sum of the denominators for those hospitals to generate the measure for the entire HRR or state.

- We downloaded the PQI software from the AHRQ website and applied it to inpatient claims. The software generates results by metropolitan statistical area; we then followed procedures developed by AHRQ to convert those results to the ZIP code level. We then added the results for all ZIP codes in each HRR or state.

We used AHRQ's software to calculate each PQI measure separately for beneficiaries under age 65, those between the ages of 65 and 74, and those who were 75 or older (with some exceptions if the measure specifications dictated otherwise; see Appendix 2).

8. Changes from the January 2013 data set to the May 2013 update

In May 2013, CMS posted a GV PUF with data for calendar years 2007-2011. This December 2013 update includes county-level data for calendar years 2008-2012 and reflects seven revisions that we have made to our methodology. The following list provides an outline of the methodological changes:

1. Changed our inclusion criteria for Part A (i.e., inpatient, LTCH, IRF, SNF, hospice, and home health) claims. We include a claim in our files if it meets three criteria: (a) the claim is a final action claim (meaning it is the version of the claim where all adjustments have been resolved), (b) the claim is linked to a beneficiary that appears in the beneficiary file for that year, and (c) the Medicare payment amount for the claim is \$0 or more. However, we have now identified some situations in the Part A claims data where we were including two versions of the same claim. For the December 2013 update we have excluded the "interim" version of a claim in the following situations: (a) when there are two or more claims that have the same beneficiary, the same provider, and either the same start date or the same end date (we kept the claim that was processed last and discarded the other(s)), (b) any interim versions of a home health claim, and (c) any home health claim that was denied, which is based on the claim having either 0 visits or a \$0 payment. The change resulted in very minor reductions to the utilization and spending measures for Part A services.

2. Modified the stay definition for Part A claims. In some cases, when a beneficiary receives care for an extended period of time, the provider will submit claims on a periodic (usually monthly) basis rather than waiting until the beneficiary is discharged. To count covered stays, we link these claims into one stay rather than counting each individual claim as a stay for a given beneficiary. In past versions of the file, we used the discharge date variable to determine when a stay ended. However, upon further investigation of the data, we have identified that using discharge date did not account for transfers among facilities or instances where a beneficiary stopped receiving a service for a period of time but was expected to return and thus not formally discharged (such as a beneficiary in a SNF who briefly returns to the hospital). As a result, we have modified our methodology to use the dates of service to link claims into a single stay. If the

dates of service are continuous or if two claims within a service category have overlapping dates, we will now count this as a single stay.

3. *Revised service categories.* In the May 2013 dataset, we had a single service category that combined inpatient rehabilitation facilities (IRFs) and long-term care hospitals (LTCHs). Due to the large volume of requests for IRF- and LTCH-specific figures, we have separated them into distinct categories for this update. In addition, in the May 2013 dataset we eliminated the service category for outpatient dialysis facilities and included these claims in an “other Part B service” category. For the December 2013 update, we brought back outpatient dialysis facility as a distinct service category and removed the “other Part B service” category.

4. *Used an updated version of the PQI software for inpatient hospital claims and used all diagnosis codes.* In the May 2013 version of the dataset, we used version 4.3 of the PQI software. For the December 2013 update, we used a newer version of the PQI software (version 4.5, which was updated in May 2013). We also used all 25 diagnosis codes on the inpatient claim, instead of just the first 10 codes.

5. *Used the original claim through date for DME claims.* Medicare allows certain DME claims, such as rentals, to be processed before the end date on the claim has been reached. In the past, if a DME claim was processed in one year but had an end date (i.e., claim through date) in the following year, we recoded the claim through date to December 31 of the year in which the claim was processed. However, in the December 2013 update, we decided to use the original claim through date for all DME claims.

6. *Updated BETOS classifications.* A recent update to the assignment of individual service codes to BETOS service categories resulted in very small changes to spending and service utilization across non-institutional Part B service categories.

7. *Addressed small inconsistencies in payments for certain Part B institutional claims.* In the December 2013 version, we edited the claim payment amount on institutional Part B claims to reconcile small inconsistencies between the sum of the payment amounts for each individual service on a claim and the total claim payment amount. This led to very small changes in actual and standardized costs for outpatient hospitals, clinics, and outpatient dialysis facilities.

Appendix 1 - Hospital Referral Regions

We list HRRs by state and the name of the primary city or county within each HRR. For maps that show the specific boundaries for each HRR, please go to:

<http://www.dartmouthatlas.org/downloads/methods/geogappdx.pdf>.

Alabama (6)	Birmingham, Dothan, Huntsville, Mobile, Montgomery, Tuscaloosa
Alaska (1)	Anchorage
Arizona (4)	Mesa, Phoenix, Sun City, Tucson
Arkansas (5)	Fort Smith, Jonesboro, Little Rock, Springdale, Texarkana
California (24)	Alameda County, Bakersfield, Chico, Contra Costa County, Fresno, Los Angeles, Modesto, Napa, Orange County, Palm Springs, Redding, Sacramento, Salinas, San Bernadino, San Diego, San Francisco, San Jose, San Luis Obispo, San Mateo County, Santa Barbara, Santa Cruz, Santa Rosa, Stockton, Ventura
Colorado (7)	Boulder, Colorado Springs, Denver, Fort Collins, Grand Junction, Greeley, Pueblo
Connecticut (3)	Bridgeport, Hartford, New Haven
Delaware (1)	Wilmington
District of Columbia (1)	Washington
Florida (18)	Bradenton, Clearwater, Fort Lauderdale, Fort Myers, Gainesville, Hudson, Jacksonville, Lakeland, Miami, Ocala, Orlando, Ormond Beach, Panama City, Pensacola, Sarasota, St. Petersburg, Tallahassee, Tampa
Georgia (7)	Albany, Atlanta, Augusta, Columbus, Macon, Rome, Savannah
Hawaii (1)	Honolulu
Idaho (2)	Boise, Idaho Falls
Illinois (13)	Aurora, Bloomington, Blue Island, Chicago, Elgin, Evanston, Hinsdale, Joliet, Melrose Park, Peoria, Rockford, Springfield, Urbana
Indiana (9)	Evansville, Fort Wayne, Gary, Indianapolis, Lafayette, Muncie, Munster, South Bend, Terre Haute
Iowa (8)	Cedar Rapids, Davenport, Des Moines, Dubuque, Iowa City, Mason City, Sioux City, Waterloo
Kansas (2)	Topeka, Wichita
Kentucky (5)	Covington, Lexington, Louisville, Owensboro, Paducah
Louisiana (10)	Alexandria, Baton Rouge, Houma, Lafayette, Lake Charles, Metairie, Monroe, New Orleans, Shreveport, Slidell
Maine (2)	Bangor, Portland
Maryland (3)	Baltimore, Salisbury, Takoma Park
Massachusetts (3)	Boston, Springfield, Worcester
Michigan (15)	Ann Arbor, Dearborn, Detroit, Flint, Grand Rapids, Kalamazoo, Lansing, Marquette, Muskegon, Petoskey, Pontiac, Royal Oak, Saginaw, St. Joseph, Traverse City

Minnesota (5)	Duluth, Minneapolis, Rochester, St. Cloud, St. Paul
Mississippi (6)	Gulfport, Hattiesburg, Jackson, Meridian, Oxford, Tupelo
Missouri (6)	Cape Girardeau, Columbia, Joplin, Kansas City, Springfield, St. Louis
Montana (3)	Billings, Great Falls, Missoula
Nebraska (2)	Lincoln, Omaha
Nevada (2)	Las Vegas, Reno
New Hampshire (2)	Lebanon, Manchester
New Jersey (7)	Camden, Hackensack, Morristown, New Brunswick, Newark, Paterson, Ridgewood
New York (10)	Albany, Binghamton, Bronx, Buffalo, East Long Island, Elmira, Manhattan, Rochester, Syracuse, White Plains
New Mexico (1)	Albuquerque
North Carolina (9)	Asheville, Charlotte, Durham, Greensboro, Greenville, Hickory, Raleigh, Wilmington, Winston-Salem
North Dakota (4)	Bismarck, Fargo, Grand Forks, Minot
Ohio (10)	Akron, Canton, Cincinnati, Cleveland, Columbus, Dayton, Elyria, Kettering, Toledo, Youngstown
Oklahoma (3)	Lawton, Oklahoma City, Tulsa
Oregon (5)	Bend, Eugene, Medford, Portland, Salem
Pennsylvania (14)	Allentown, Altoona, Danville, Erie, Harrisburg, Johnstown, Lancaster, Philadelphia, Pittsburgh, Reading, Sayre, Scranton, Wilkes-Barre, York
Rhode Island (1)	Providence
South Carolina (5)	Charleston, Columbia, Florence, Greenville, Spartanburg
South Dakota (2)	Rapid City, Sioux Falls
Tennessee (7)	Chattanooga, Jackson, Johnson City, Kingsport, Knoxville, Memphis, Nashville
Texas (22)	Abilene, Amarillo, Austin, Beaumont, Bryan, Corpus Christi, Dallas, El Paso, Fort Worth, Harlingen, Houston, Longview, Lubbock, McAllen, Odessa, San Angelo, San Antonio, Temple, Tyler, Victoria, Waco, Wichita Falls
Utah (3)	Ogden, Provo, Salt Lake City
Vermont (1)	Burlington
Virginia (8)	Arlington, Charlottesville, Lynchburg, Newport News, Norfolk, Richmond, Roanoke, Winchester
West Virginia (3)	Charleston, Huntington, Morgantown
Wisconsin (8)	Appleton, Green Bay, La Crosse, Madison, Marshfield, Milwaukee, Neenah, Wausau
Washington (6)	Everett, Olympia, Seattle, Spokane, Tacoma, Yakima
Wyoming (1)	Casper

Appendix 2 – Quality Measures Included in the GV PUF

Hospital Compare (30 measures, calculated per 100 patients)

Heart attack patients given aspirin at arrival

Heart attack patients prescribed aspirin at discharge

Heart attack patients given ACE inhibitor or ARB for LVSD

Heart attack patients given smoking cessation advice / counseling

Heart attack patients given beta blocker at discharge

Heart attack patients given fibrinolytic medication within 30 minutes of arrival

Heart attack patients given PCI within 90 minutes of arrival

30-day death rate for heart attack patients

Hospital 30-day readmission rate for heart attack patients

Heart failure patients given discharge instructions

Heart failure patients given an evaluation of left ventricular systolic function

Heart failure patients given ACE inhibitor or ARB for LVSD

Heart failure patients given smoking cessation advice / counseling

30-day death rate for heart failure patients

Hospital 30-day readmission rate for heart failure patients

Pneumonia patients assessed and given pneumococcal vaccination

Pneumonia patients with initial ER blood culture performed prior to initial antibiotic in hospital

Pneumonia patients given smoking cessation advice / counseling

Pneumonia patients given initial antibiotic(s) within 6 hours of arrival

Pneumonia patients given the most appropriate initial antibiotic(s)

Pneumonia patients assessed and given influenza vaccination

30-day death rate for pneumonia patients

Hospital 30-day readmission rate for pneumonia patients

Surgery patients received preventative antibiotic(s) 1 hour before incision

Surgery patients received the appropriate preventative antibiotic(s) for their surgery

Surgery patients had preventative antibiotic(s) stopped within 24 hours after surgery

Cardiac surgery patients with controlled 6 AM postoperative blood glucose

Surgery patients with appropriate hair removal

Surgery patients whose doctors ordered VTE for certain types of surgeries

Surgery patients who received appropriate VTE within 24 hours before or after certain surgeries

Prevention Quality Indicators (9 measures, calculated per 100,000 beneficiaries in the specified age groups)

Diabetes long-term complications admission rate (<65, 65-74, 75+)

Chronic obstructive pulmonary disease or asthma in older adults admission rate (40-64, 65-74, 75+)

Hypertension admission rate (<65, 65-74, 75+)

Congestive heart failure admission rate (<65, 65-74, 75+)

Dehydration admission rate (<65, 65-74, 75+)

Prevention Quality Indicators, continued

Bacterial pneumonia admission rate (<65, 65-74, 75+)

Urinary tract infection admission rate

Asthma in younger adults (<40)

Rate of lower extremity amputations among patients with diabetes (<65, 65-74, 75+)

Readmissions and Emergency Room Use (4 measures)

Total number of hospital readmissions

Hospital readmission rate

Total number of emergency room visits

Total number of emergency room visits per 1000 beneficiaries

Abbreviations: ACE = angiotensin-converting enzyme, ARB = angiotensin receptor blocker, ER = emergency room, LVSD = left ventricular systolic dysfunction, PCI = percutaneous coronary intervention, VTE = venous thromboembolism