

Respiratory Infection Hospitalization

Measure Justification Form

December 2023



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1.0 Introduction

This Measure Justification Form (MJF) provides results for the testing and evaluation of the Respiratory Infection Hospitalization measure. The form is intended to provide detailed information about the testing conducted on this measure, and accompanies the Measure Methodology¹ and Measure Codes List² file, which together, comprise the specifications for this cost measure.

1.1 Project Title

Physician Cost Measure and Patient Relationship Codes

1.2 Date

Information included is current on December 8, 2023

1.3 Project Overview

The Centers for Medicare & Medicaid Services (CMS) has contracted with Acumen, LLC to develop care episode and patient condition groups for use in cost measures to meet the Medicare Access and CHIP Reauthorization Act of 2015 (MACRA) requirements. The contract name is "Physician Cost Measure and Patient Relationship Codes (PCMP)." The contract number is 75FCMC18D0015, Task Order 75FCMC19F0004.

1.4 Measure Name

Respiratory Infection Hospitalization Episode-Based Cost Measure

1.5 Type of Measure

Cost/Resource Use

1.6 Measure Description

Episode-based cost measures represent the cost to Medicare for the items and services provided to a patient during an episode of care ("episode"). In all supplemental documentation, the term "cost" generally means the standardized³ Medicare allowed amount⁴, and claims data from Medicare Parts A and B⁵ are used to construct this episode-based cost measure.

¹CMS, "Respiratory Infection Hospitalization Measure Methodology," *QPP Cost Measure Information Page*, <https://www.cms.gov/medicare/quality/value-based-programs/cost-measures>

²CMS, "Respiratory Infection Hospitalization Measure Codes List" *QPP Cost Measure Information Page*, <https://www.cms.gov/medicare/quality/value-based-programs/cost-measures>

³ Claim payments are standardized to account for differences in Medicare payments for the same service(s) across Medicare providers. Payment standardized costs remove the effect of differences in Medicare payment among health care providers that are the result of differences in regional health care provider expenses measured by hospital wage indexes and geographic price cost indexes or other payment adjustments such as those for teaching hospitals. For more information, please refer to the "CMS Part A and Part B Price (Payment) Standardization - Basics" and "CMS Part A and Part B Price (Payment) Standardization - Detailed Methods" documents posted on the [CMS Price \(Payment\) Standardization Overview](https://www.resdac.org/articles/cms-price-payment-standardization-overview) page (<https://www.resdac.org/articles/cms-price-payment-standardization-overview>).

⁴ Cost is defined by allowed amounts on Medicare claims data, which include both Medicare trust fund payments and any applicable beneficiary deductible and coinsurance amounts.

⁵ Part D branded drug costs are also adjusted to account for post-point of sale drug rebates; more information can be found in the [Methodology for Rebates in Part D Standardized Amounts](https://www.cms.gov/medicare/quality-payment-program/cost-measures/about) on the CMS.gov QPP Cost Measures Information Page's [QPP Cost Measure Information page](https://www.cms.gov/medicare/quality-payment-program/cost-measures/about) (<https://www.cms.gov/medicare/quality-payment-program/cost-measures/about>).

The Respiratory Infection Hospitalization episode-based cost measure evaluates a clinician's or clinician group's risk-adjusted cost to Medicare for patients who receive inpatient treatment for a respiratory infection. This acute measure includes the costs of services that are clinically related to the attributed clinician's role in managing care during a Respiratory Infection Hospitalization episode.

2.0 Importance

2.1 Evidence to Support the Measure Focus

The Respiratory Infection Hospitalization measure was developed for use in the Merit-based Incentive Payment System (MIPS) to meet the requirements of the Social Security Act section 1848(r), added by the Medicare Access and CHIP Reauthorization Act of 2015 (MACRA). MIPS aims to reward high-value care by measuring clinician performance through four areas: quality, improvement activities, promoting interoperability, and cost. Each category assesses different aspects of care, and the categories are weighted to combine into one composite score. CMS is introducing MIPS Value Pathways (MVPs) to align and connect quality measures, cost measures, and improvement activities across performance categories of MIPS for different specialties or conditions. MVPs aim to provide a holistic assessment of clinician value for a specific type of care to achieve better healthcare outcomes and lower patient costs.

The use of cost measures is required by statute, and their purpose is to assess resource use. To be effective, they should capture costs related to a clinician's care decisions and account for factors outside their influence. This measure provides clinicians with information about their care costs that they can use to understand the costs associated with their decision-making. Clinicians play an important role in variation in health care expenditures due to their ability to affect costs.⁶ A cost measure offers an opportunity for improvement if clinicians can exercise influence on the intensity or frequency of a significant share of costs during the episode, or if clinicians can achieve lower spending and better quality of care quality through changes in clinical practice.

According to the literature and feedback received through stakeholder input activities, this measure's focus represents an area with opportunities for improvement. As discussed in the rest of this section, primary opportunities for improving respiratory infection hospitalization cost outcomes include reducing hospital readmissions and overuse of antibiotics.

Respiratory infections are a leading cause of mortality and morbidity; 85% of recorded deaths in the United States are due to respiratory-tract infections, and over \$16.1 billion was spent on respiratory-tract infections in 2013.⁷ Pneumonia is a respiratory tract infection that constitutes a substantial disease burden in adults, especially the elderly. In 2015, nearly 6.8 million episodes of clinical pneumonia resulted in hospital admissions in older adults.⁸ According to the CDC, pneumonia and influenza were the 10th leading causes of mortality in adults aged ≥65 in the United States in 2020.⁹ Further, Community Acquired Pneumonia (CAP), is four times more

⁶David Cutler et al., "Physician Beliefs and Patient Preferences: A New Look at Regional Variation in Health Care Spending," *American Economic Journal: Economic Policy* 11, no. 1 (February 1, 2019): 192–221, <https://doi.org/10.1257/pol.20150421>.

⁷ Akhtar, Ali, Azmi Ahmad Hassali Mohamed, Zainal Hadzliana, Ali Irfhan, Muhammad Shahid Iqbal, and Amer Hayat Khan. "Respiratory-Tract Infections among Geriatrics: Prevalence and Factors Associated with the Treatment Outcomes." *Therapeutic Advances in Respiratory Disease* 15, (01, 2021). <https://doi.org/10.1177/1753466620971141>.

⁸ Shi, Ting, Angeline Denouel, Anna K. Tietjen, Jen Wei Lee, Ann R. Falsey, Clarisse Demont, Bryan O. Nyawanda, et al. "Global and Regional Burden of Hospital Admissions for Pneumonia in Older Adults: A Systematic Review and Meta-Analysis." *The Journal of Infectious Diseases* 222, Oct 07, 2020. Pages S570–S576, <https://doi.org/10.1093/infdis/jiz053>.

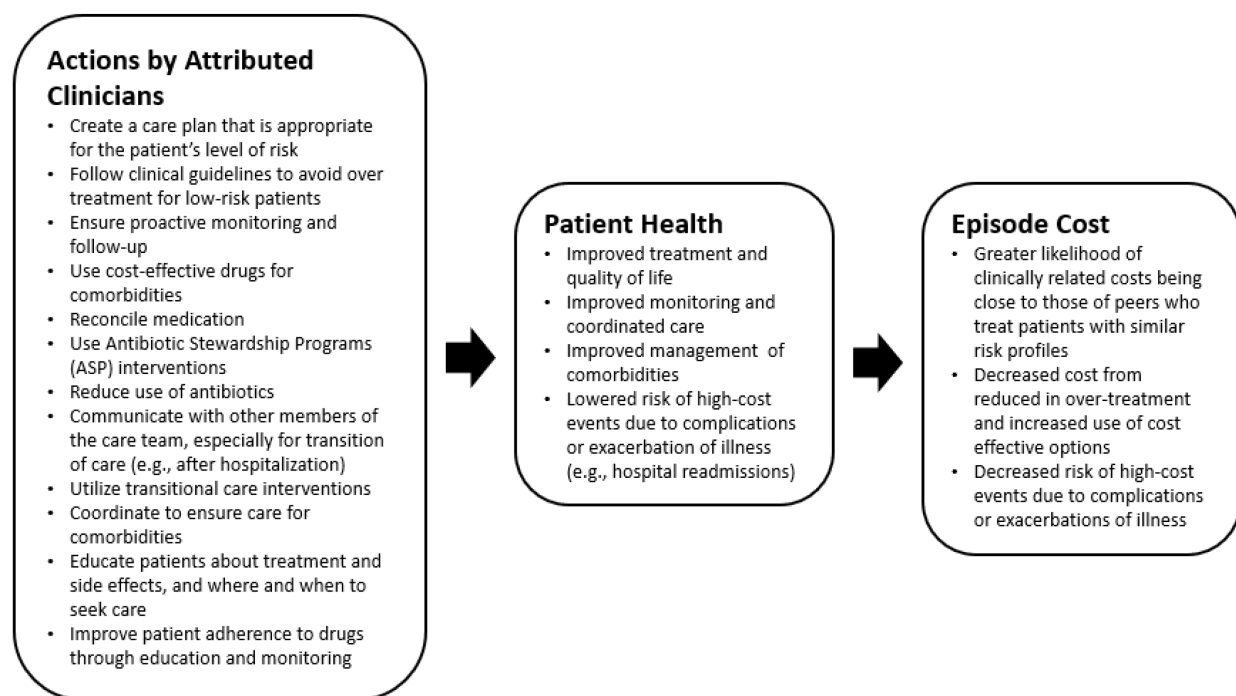
⁹ Centers for Disease Control and Prevention. WISQARS. Leading Causes of Death Visualization Tool. <https://wisqars.cdc.gov/data/lcd/home>.

likely to occur in this age group than their younger counterparts, and the risk of hospitalization is also higher.¹⁰

Research shows that CAP poses a serious clinical and cost burden for the elderly, drawing attention to the need to develop effective prevention and treatment strategies for this population. One study among the older US population found that when compared to diabetes mellitus (DM), myocardial infarction (MI), and stroke, the primary payer spent the most on CAP hospitalizations alone, \$1,130 (in millions), which exceeded the combined cost of DM and stroke hospitalizations. According to the study, this difference was due to hospitalization incidence for CAP being roughly double that of MI and stroke and even larger compared to DM.¹¹ CAP also is attributed with long-term clinical costs. A study examining expenditures of CAP episodes requiring hospitalizations found that about 21%, 14%, and 12% of CAP-related expenditures occurred during the first, second, and third years, respectively, after 30 days post-discharge.¹²

2.1.1 Logic Model

Figure 1: Logic Model of Steps between Actions by Attributed Clinicians and Episode Cost



¹⁰ Olasupo, Omotola, Hong Xiao, and Joshua D. Brown. "Relative Clinical and Cost Burden of Community-Acquired Pneumonia Hospitalizations in Older Adults in the United States—A Cross-Sectional Analysis." *Vaccines* 6, no. 3 (09, 2018). <https://doi.org/10.3390/vaccines6030059>.

¹¹ Olasupo, Omotola, Hong Xiao, and Joshua D. Brown. "Relative Clinical and Cost Burden of Community-Acquired Pneumonia Hospitalizations in Older Adults in the United States—A Cross-Sectional Analysis." *Vaccines* 6, no. 3 (09, 2018). <https://doi.org/10.3390/vaccines6030059>.

¹² Weycker, Derek, Aaron Moynahan, Amanda Silvia, and Reiko Sato. "Attributable Cost of Adult Hospitalized Pneumonia Beyond the Acute Phase." *PharmacoEconomics - Open* 5, no. 2 (06, 2021): 275-284. <https://doi.org/10.1007/s41669-020-00240-9>.

2.2 Performance Gap

2.2.1 Rationale

Pneumonia is responsible for nearly 140,000 readmissions annually, and more than \$10 billion in hospital expenditures. Additionally, the number of pneumonia-related readmissions has increased significantly in recent years, as a result of an aging population, antibiotic resistance patterns and an increasing prevalence of comorbidities. Readmissions not only impose an additional burden onto vulnerable populations who are most commonly affected by pneumonia, but are also costly.¹³ One report predicted that hospital readmissions within 30 days cost 3.3 million, and affected more than 55% of patients on Medicare in 2011.¹⁴ Transitional care interventions can reduce risk of readmission among beneficiaries hospitalized with a respiratory infection. For instance, a pilot program launched by Stanford University Medical Center found that a transitional program including patient education and home visits for the most at-risk patients reduced pneumonia readmission rates from 17.4% to 11.5%.¹⁵ Additionally, evidence has shown that use of standard guidelines to care for Medicare beneficiaries hospitalized with pneumonia can decrease readmission rates.¹⁶ Overall, efforts to reduce readmission rates are not only important to improve patient outcomes, but also healthcare spending as the average cost of readmissions for adults aged 65 years or older is \$15,976.¹⁷

Moreover, community-onset infections of any kind and lower respiratory tract infections acquired in any setting are the most common indicators for antibiotic use among hospitalized patients. Previous studies have affirmed that the average length of antibiotic therapy for CAP often exceeds national recommendations set forth by the Infectious Diseases Society of America (IDSA) and the American Thoracic Society (ATS). Adults hospitalized for CAP in the United States received a median of just under 10 days of antibiotic therapy in 2012–2013 with more than 70% of patients exceeding the recommended duration of antibiotics.¹⁸ Additionally, a retrospective analysis of medical records found that in a sample of elderly patients with multiple comorbidities hospitalized for CAP, 52% of the population met the ATS/ISDA clinical stability criteria at day 5, highlighting the effectiveness of shorter antibiotic courses.¹⁹ Furthermore, the use of Antibiotic Stewardship Programs (ASP) has demonstrated the ability to reduce antibiotic use. In a study in which adult patients aged 65 and older who received antibiotic treatments during an inpatient stay were subject to ASP interventions, those with a pneumonia (PNA)

¹³ Alba, Israel De and Alpesh Amin. "Pneumonia Readmissions: Risk Factors and Implications." *The Ochsner Journal* 14, no. 4 (12, 2014): 649-654. <https://www.proquest.com/scholarly-journals/pneumonia-readmissions-risk-factors-implications/docview/2157950821/se-2>.

¹⁴ Flanagan, Jane, Kelly D. Stamp, Matt Gregas, and Judy Shindul-Rothschild. "Predictors of 30-Day readmission for pneumonia." *The Journal of Nursing Administration* 46, no. 2 (2016): 69-74.

¹⁵ Transitional Care Reduces Pneumonia Readmissions. Stanford Health Care. November 26, 2016. Accessed May 18, 2023. <https://stanfordhealthcare.org/content/dam/SHC/clinics/aging-adult-services/docs/10.31.16%20MedStaff%20Update.pdf>.

¹⁶ Dean NC, Bateman KA, Donnelly SM, Silver MP, Snow GL, Hale D. Improved clinical outcomes with utilization of a community-acquired pneumonia guideline. *Chest*. 2006;130(3):794-799. doi:10.1378/chest.130.3.794.

¹⁷ Jain S, Khera R, Mortensen EM, Weissler JC. Readmissions of adults within three age groups following hospitalization for pneumonia: Analysis from the Nationwide Readmissions Database. *PLoS One*. 2018;13(9):e0203375. Published 2018 Sep 13. doi:10.1371/journal.pone.0203375

¹⁸ Yi, Sarah H., Kelly M. Hatfield, James Baggs, Lauri A. Hicks, Arjun Srinivasan, Sujan Reddy, and John A. Jernigan. "Duration of Antibiotic use among Adults with Uncomplicated Community-Acquired Pneumonia Requiring Hospitalization in the United States." *Clinical Infectious Diseases : An Official Publication of the Infectious Diseases Society of America* 66, no. 9 (Apr 17, 2018): 1333-1341. <https://doi.org/10.1093/cid/cix986>.

¹⁹ Flateau, C., M. Dinia, N. Raulet, S. Sayegh, S. Diamantis, and M. Jager. "Does a 5-Day Course of Antibiotics in Elderly Patients with Community-Acquired Pneumonia Achieve the Established Criteria of Clinical Stability?" *Infectious Diseases Now* 51, no. 4 (06, 2021): 377-379. <https://doi.org/10.1016/j.medmal.2020.10.015>.

diagnosis exhibited a significant reduction in readmission rates along with a significant decrease in antibiotic expenditure.²⁰ ASP interventions supported by the use of biomarkers have similarly demonstrated success in reinforcing the appropriate use of antibiotics. Specifically, in a study of hospitalized patients with LTRIs the use of Procalcitonin (PCT), a biomarker that has shown promising results in guiding antibiotic therapy,²¹ LRTIs reduced total costs by \$2,867, a difference driven by a reduction in patient length of stay and antibiotic resistance.²²

The Simple Pneumonia with Hospitalization episode-based measure was originally developed because of its high impact in terms of patient population and Medicare spending. The revised Respiratory Infection Hospitalization episode-based measure increases the number of clinicians participating in the measure without compromising the measure's reliability. This was achieved by expanding the patient cohort to include beneficiaries with respiratory infections and inflammations. This subgroup was added to the measure based on input from the Clinician Expert Workgroup, and expected cost differences not under the influence of the attributed clinician are accounted for through risk adjustment. Further, as evidenced by the literature review, there are opportunities to improve efficiency (i.e., reduce overuse of antibiotics and hospital readmissions) thereby reduce cost to Medicare for patients hospitalized with a respiratory infection.

2.2.2 Performance Scores

Table 1 shows the distribution of the measure score for clinician groups identified by a Tax Identification Number (TIN) and individual clinicians identified by a combination of a Tax Identification Number and National Provider Identifier (TIN-NPI).

There are variations in cost performance observed in the measure score for both TINs and TIN-NPIs, as evidenced by the interquartile ranges and score standard deviations. For both TINs and TIN-NPIs, the maximum score is about 2 times larger than the minimum score. The variation in the measure score is in the thousands of dollars, which highlights an opportunity for improvement in the costs of care for a respiratory infection hospitalization episode by closing the gap between the most and least efficient providers.

Table 1. Distribution of the Measure Score

Metric	TIN	TIN-NPI
Count	3,169	10,254
Mean Score	\$15,066	\$17,207
Score Standard Deviation	\$1,352	\$1,893
Minimum Score	\$10,423	\$12,171
Maximum Score	\$21,670	\$25,831
Score Interquartile Range (IQR)	\$1,582	\$2,505
Score Percentile		

²⁰ Mauro, James, Saman Kannangara, Joanne Peterson, David Livert, and Roman A. Tuma. "Rigorous Antibiotic Stewardship in the Hospitalized Elderly Population: Saving Lives and Decreasing Cost of Inpatient Care." JAC-Antimicrobial Resistance 3, no. 3 (09, 2021): 1. <https://doi.org/10.1093/jacamr/dlab118>.

²¹ Falcone, Marco, Michael Bauer, Ricard Ferrer, Gaëtan Gavazzi, Juan Gonzalez Del Castillo, Alberto Pilotto, and Philipp Schuetz. "Biomarkers for Risk Stratification and Antibiotic Stewardship in Elderly Patients." Aging Clinical and Experimental Research (Mar 30, 2023). <https://doi.org/10.1007/s40520-023-02388-w>.

²² Mewes, Janne C., Michael S. Pulia, Michael K. Mansour, Michael R. Broyles, H. B. Nguyen, and Lotte M. Steuten. "The Cost Impact of PCT-Guided Antibiotic Stewardship Versus Usual Care for Hospitalised Patients with Suspected Sepsis Or Lower Respiratory Tract Infections in the US: A Health Economic Model Analysis." PloS One 14, no. 4 (2019): 1. <https://doi.org/10.1371/journal.pone.0214222>.

Metric	TIN	TIN-NPI
10 th	\$13,527	\$14,917
20 th	\$14,024	\$15,588
30 th	\$14,359	\$16,136
40 th	\$14,651	\$16,590
50 th	\$14,959	\$17,061
60 th	\$15,268	\$17,535
70 th	\$15,601	\$18,066
80 th	\$16,032	\$18,730
90 th	\$16,766	\$19,739

2.2.3 Disparities

Data on how the measure, as specified, addresses disparities is described in Sections 3.1.7 and 3.5.5.

3.0 Scientific Acceptability

3.1 Data Sample Description

Testing is based on the full population of measured entities and patients meeting inclusion and exclusion criteria for the measure, not based on a sample.

3.1.1 Type of Data Used for Testing

Medicare administrative claims data from the Common Working File (CWF), Long-Term Care Minimum Data Set (LTC MDS), and Medicare Enrollment Database (EDB).

3.1.2 Specific Dataset Used for Testing

The Respiratory Infection Hospitalization measure uses Medicare Part A and Part B claims data maintained by CMS. Part A and B claims data are used to build episodes of care, calculate episode costs, and construct risk adjusters. Episode costs are payment standardized and risk adjusted to ensure accurate comparison of cost across clinicians. Payment standardization adjusts the allowed amount for a Medicare service to limit observed differences in costs to those that may result from health care delivery choices. Data from the EDB are used to determine beneficiary-level exclusions and secondary risk adjusters, specifically Medicare Parts A, B, and C enrollment, primary payer, disability status, end-stage renal disease (ESRD), patient birth dates, and patient death dates. The risk adjustment model also accounts for expected differences in payment for services provided to patients in long-term care based on data from the LTC MDS. Specifically, the LTC MDS is used to create the long-term care indicator variable in risk adjustment.

3.1.3 Dates of the Data Used in Testing

Respiratory Infection Hospitalization episodes ending from January 1, 2022, through December 31, 2022.

3.1.4 Levels of Analysis Tested

The measure was tested at group/practice (TIN) and individual clinician (TIN-NPI) levels.

3.1.5 Entities Included in the Testing and Analysis

Table 2 shows the individual clinician (identified by combination of TIN and NPI) and clinician group/practice (identified by TIN) included in the testing of the Respiratory Infection Hospitalization measure.

Table 2: Measured Entities Demographics

Metric	TIN		TIN-NPI	
	Count	%	Count	%
Count	3,169	100.00%	10,254	100.00%
Number of Episodes Attributed	-	-	-	-
20-39 Episodes	1,130	35.66%	9,083	88.58%
40-59 Episodes	532	16.79%	902	8.80%
60-79 Episodes	302	9.53%	174	1.70%
80-99 Episodes	215	6.78%	50	0.49%
100-199 Episodes	535	16.88%	45	0.44%
200-299 Episodes	203	6.41%	0	0.00%
300+ Episodes	252	7.95%	0	0.00%
Census Region	-	-	-	-

Metric	TIN		TIN-NPI	
	Count	%	Count	%
Northeast	537	16.95%	2,614	25.49%
Midwest	768	24.23%	2,248	21.92%
South	1,340	42.28%	4,367	42.59%
West	521	16.44%	1,024	9.99%
Unknown	3	0.09%	1	0.01%

3.1.6 Patient Cohort Included in the Testing and Analysis

Table 3 shows the patient population for the Respiratory Infection Hospitalization measure testing. It consists of Medicare beneficiaries enrolled in Medicare Parts A and B who receive inpatient treatment for a respiratory infection that triggers a Respiratory Infection Hospitalization episode and do not meet the measure's exclusion criteria, as outlined in section 3.4.1.

Table 3: Beneficiary Demographics

Metric	Value
Count	328,463
Mean Age	77.35 years
Female %	53.52%

3.1.7 Social Risk Factors Included in Analysis

The analysis of social risk factors (SRFs) focused on examining the impact of Dual Medicare and Medicaid enrollment status on the measure. Table 4 outlines variables that may indicate SRFs and their advantages and disadvantages as indicators of individual-level SRFs. On balance, the analysis used dual Medicare and Medicaid enrollment status as the proxy of SRFs due to their broad availability in claims data, accurate measurement at the individual level, and wide acceptance of being a powerful indicator of health outcomes.²³

Table 4: Social Risk Factors Available for Analysis

Variable	Advantages	Disadvantages	Used in Testing
Dual Medicare and Medicaid enrollment status	<ul style="list-style-type: none"> Available for all beneficiaries Most powerful predictor of poor outcomes²³ 	<ul style="list-style-type: none"> Variation in Medicaid eligibility across states 	Yes
Race/Ethnicity	<ul style="list-style-type: none"> Available for most beneficiaries, except for ambiguous categories of "Unknown" or "Other" 	<ul style="list-style-type: none"> Social risk driven by someone's race is often correlated with and partially captured by dual status²³ Only 5 categories available, which may lack granularity 	No

²³ Office of the Assistant Secretary for Planning and Evaluation. "Second report to Congress on social risk and Medicare's value-based purchasing programs." (2020) <https://aspe.hhs.gov/pdf-report/second-impact-report-to-congress>

Variable	Advantages	Disadvantages	Used in Testing
		to fully capture disparities ^{24,25}	
ICD-10 Z codes for social determinants of health	<ul style="list-style-type: none"> Reflects individual-level factors that influence health status and contact with health services 	<ul style="list-style-type: none"> Not routinely and consistently coded on claims, only available for 0.1% of all fee-for-service claims in 2019²⁶ 	No
American Community Survey	<ul style="list-style-type: none"> Can link beneficiary's zip code to socioeconomic (SES) measurement of their neighborhood Many SES indices can be derived from the survey data (e.g., AHRQ index, deprivation index) 	<ul style="list-style-type: none"> Only a proxy measure, not always accurate at individual-level 	No

3.2 Reliability Testing

3.2.1 Level of Reliability Testing

The following levels of reliability were tested: critical data elements used in the measure, group/practice (TIN) and individual clinician (TIN-NPI) levels.

3.2.2 Method of Reliability Testing

Data Element Reliability

The Respiratory Infection Hospitalization measure is constructed using CMS claims data, as described in Section 3.1.2. CMS has implemented several auditing programs to assess overall claims code accuracy, ensure appropriate billing, and recoup any overpayments.

- First, CMS routinely conducts data analyses to identify potential problem areas and detect fraud and audits necessary data fields used in this measure, including diagnosis and procedure codes and other elements consequential to payment. Specifically, CMS works with Zone Program Integrity Contractors, formerly Program Safeguard Contractors, to ensure program integrity; the agency also uses Recovery Audit Contractors to identify and correct for underpayments and overpayments.
- Second, CMS also uses the Comprehensive Error Rate Testing (CERT) Program to ensure that Medicare payments are correct under coverage, coding, and billing rules. CMS continues to perform corrective actions and give providers additional education to ensure accurate billing.

²⁴ Nguyen, Kevin H., Kaitlyn P. Lew, and Amal N. Trivedi. "Trends in Collection of Disaggregated Asian American, Native Hawaiian, and Pacific Islander Data: Opportunities in Federal Health Surveys." *American Journal of Public Health* (2022).

²⁵ Kader, Farah, Lan N. Doan, Matthew Lee, Matthew K. Chin, Simona C. Kwon, and Stella S. Yi. "Disaggregating Race/Ethnicity Data Categories: Criticisms, Dangers, And Opposing Viewpoints", *Health Affairs Forefront* (2022).

²⁶ Centers for Medicare and Medicaid, Office of Minority Health. "Utilization of Z Codes for Social Determinants of Health among Medicare Fee-for-Service Beneficiaries." (2019) <https://www.cms.gov/files/document/z-codes-data-highlight.pdf>

- Lastly, to ensure claims completeness and inclusion of any corrections, the measure was developed and tested using data with three-month claims run-out from the end of the measurement period.

Clinician-level Reliability

Measure reliability is the degree to which repeated measurements of the same entity agree with each other). For measures of clinician performance, the measured entity is the TIN or TIN-NPI, and reliability is the extent to which repeated measurements of the TIN or TIN-NPI give similar results. To estimate measure reliability, we used a signal-to-noise analysis.

This approach seeks to determine how much of the variation in the measure score is explained by differences among clinician performance (i.e., signal) rather than random variation (i.e., statistical noise) among clinicians due to the sample of cases observed. To achieve this, we calculate reliability scores as:

$$R_j = \frac{\sigma_b^2}{\sigma_b^2 + \sigma_{w_j}^2}$$

Where:

$\sigma_{w_j}^2$ is the within-group variance of the mean measure score of clinician j

σ_b^2 is the between-group variance of clinicians within the episode group

That is, reliability is calculated as the ratio of between-group variance to the sum of between-group variance and within-group variance. Reliability closer to a value of one indicates that the between-group variance is relatively large compared to the within-group variance, which suggests that the measure is effectively capturing the systematic differences between the clinician and their peer cohort.

3.2.3 Statistical Results from Reliability Testing

Data Element Reliability

Between 2005 and 2020, CMS Comprehensive Error Rate Testing (CERT) estimates that proper payment, which includes payments that met Medicare coverage, coding, and billing rules, ranged from 87.3% to 93.7% of total payments each year.²⁷ The fiscal year 2022 Medicare fee-for-service program proper payment rate was 92.5%.²⁸

Clinician-level Reliability

The table below shows reliability metrics at the 20-episode testing volume threshold. While higher thresholds generally yield higher reliability results, these increases must be considered against decreasing the number of clinicians and clinician groups eligible for the measure, which would limit the applicability of measures to larger group practices and potentially limit the impact of the measure in encouraging performance improvement. For testing purposes, we used a 20-episode volume threshold. If the measure is implemented in MIPS in the future, CMS will establish a case minimum through notice-and-comment rulemaking.

²⁷Comprehensive Error Rate Testing (CERT) Program. "Appendices Medicare Fee-for-Service 2020 Improper Payments Report". Table A6. <https://www.cms.gov/files/document/2020-medicare-fee-service-supplemental-improper-payment-data.pdf-1>.

²⁸Ibid.

Table 5: Reliability at the Accountability Entity Level

Reporting Level	Entities Meeting Case Minimum	Mean Reliability	Median Reliability	% Above 0.4	% Above 0.7
TIN	3,169	0.74	0.73	100%	56.58%
TIN-NPI	10,254	0.53	0.51	100%	3.39%

3.2.4 Interpretation

The results of the data element testing show very high reliability of the critical data elements used by the measure. At the accountability entity level, the measure is highly reliable for both the TIN and TIN-NPI reporting levels, at 0.74 and 0.53 respectively. A measure with high reliability suggests that performance comparisons across clinicians reflects systematic differences in actual performance better. Based on existing scientific evidence on the different interpretations and methods of estimating reliability, CMS finalized in the CY 2022 Physician Fee Schedule (86 FR 64996) rule that the 0.4 threshold for mean reliability continues to be appropriate for indicating moderate reliability for performance measures in the Cost category in the MIPS program. Mean reliability levels above 0.7 continue to demonstrate high reliability for cost measures, as previously established in the CY 2017 Quality Payment Program final rule (81 FR 77169 through 77171).²⁹ Additionally, at each testing volume threshold, 100% of TINs and TIN-NPIs meet or exceed the moderate reliability threshold of 0.4 and 56.58% and 3.39% of TINs and TIN-NPIs, respectively, are above the high reliability threshold of 0.7.

3.3 Validity Testing

3.3.1 Level of Validity Testing

The validity of the measure was tested using empirical validity at the accountable entity level (TIN and TIN-NPI).

3.3.2 Method of Validity Testing

Face Validity

The Respiratory Infection Hospitalization measure was developed through a structured, iterative process for gathering detailed input on the measure from recognized clinician experts. Experts in this clinical area evaluated specifications to ensure that each aspect of the measure (e.g., assigned services) was intentionally capturing only the costs of care within the reasonable influence of the attributed clinician for a defined patient population (i.e., the ability of the measure score to differentiate between good from poor performance).

In developing this measure, Acumen incorporated input from:

- (i) a Respiratory Infection Hospitalization Clinician Expert Workgroup;
- (ii) a Technical Expert Panel (TEP); and
- (iii) the Person and Family Partners.

²⁹ CMS, “Medicare Program; CY 2022 Payment Policies Under the Physician Fee Schedule and Other Changes to Part B Payment Policies; Medicare Shared Savings Program Requirements; Provider Enrollment Regulation Updates; and Provider and Supplier Prepayment and Post-Payment Medical Review Requirements,” [86 FR 64996-66031](#).

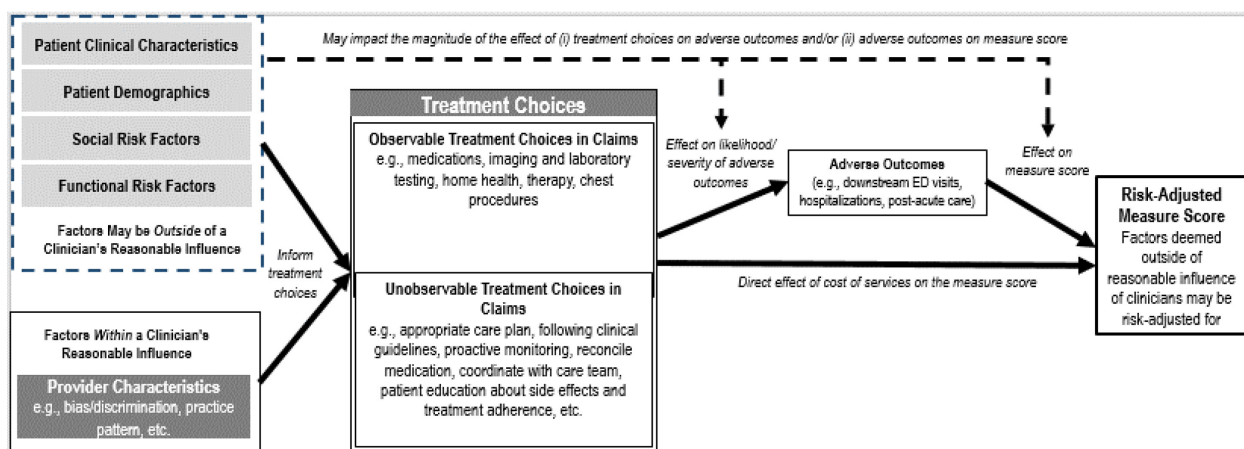
This process is detailed in the Episode-Based Cost Measures Development Process document posted on the [QPP Cost Measure Information Page](#).³⁰

One of the primary roles of the Clinician Expert Workgroup is to develop service assignment rules for the cost measure. These service assignment rules seek to ensure clinicians are evaluated on services and costs that are clinically related to the attributed clinician's role in managing the respiratory infection during hospitalization, thus limiting cost variation unrelated to clinician care in this measure. Therefore, assigned services are services that the Clinical Expert Workgroup believed an attributed clinician could influence their occurrence, frequency, or intensity.

Empirical Validity Testing

Validity is a criterion used to assess whether the cost measure can quantify the construct it aims to measure, which is the cost directly related to treatment choices and the cost of adverse outcomes resulting from care. We evaluated the empirical validity of the Respiratory Infection Hospitalization measure by estimating the effect of relevant treatment choices on the measure score using multiple regression, based on the conceptual model outlined in Figure 2.

Figure 2: Conceptual Model of Treatment Choices on the Measure Score



The cost measure is designed to reflect costs directly related to treatment choices, and the cost of adverse outcomes resulting from care. Therefore, treatment choices, either observable in claims or otherwise, by an attributed clinician can directly impact the measure score or indirectly when they are mediated through the cost of adverse outcomes. In turn, the cost of adverse effects to the total cost captured by the measure score.

This analysis first estimates the association between treatment choices and the measure score while controlling for the cost of adverse outcomes to demonstrate that the score reflects both the direct and indirect effects of treatment choices. Then, the association between treatment choices and the cost of adverse outcomes is estimated to illustrate the indirect effect.

Generally, adverse outcomes are non-trigger inpatient hospitalizations, non-trigger emergency room visits, and post-acute care. The remaining cost categories are generally considered treatment. For each of these categories, the regression models use the mean cost across episodes that were attributed to an individual clinician. The measure score is represented by a clinician's mean observed cost over expected cost ratio across their attributed episodes.

³¹CMS, QPP Cost Measure Information Page, <https://www.cms.gov/medicare/quality/value-based-programs/cost-measures>.

3.3.3 Statistical Results from Validity Testing

Empirical Validity Testing

Table 6 shows two regression models for each reporting level. Model 1 shows the effect on the clinicians' mean observed cost to expected cost ratio for each additional one thousand dollar of a cost category that is assigned to an episode, on average, while holding the remaining categories of cost constant. Model 2 shows the effect on the mean cost of adverse events for each additional one thousand dollar of a cost category that is assigned to an episode, on average, while holding the remaining categories of cost constant.

Table 6. Estimated Effect on Treatment Choices on the Measure Score

Service Categories	Coefficient in Thousands [95% Confidence Interval] (p-value)			
	TIN		TIN-NPI	
	Model 1: Mean O/E = Mean Cost of Treatment Choices + Mean Cost of Adverse Events	Model 2: Mean Cost of Adverse Events = Mean Cost of Treatment Choices	Model 1: Mean O/E = Mean Cost of Treatment Choices + Mean Cost of Adverse Events	Model 2: Mean Cost of Adverse Events = Mean Cost of Treatment Choices
Adverse Events	0.05 [0.05,0.05] (p < 0.01)	-	0.06 [0.05,0.06] (p < 0.01)	-
Outpatient Evaluation & Management Services	0.09 [0.06,0.12] (p < 0.01)	10.46 [9.87,11.05] (p < 0.01)	0.09 [0.07,0.11] (p < 0.01)	9.82 [9.53,10.12] (p < 0.01)
Ambulatory/Minor Procedures	0.22 [-0.08,0.52] (p = 0.15)	3.57 [-3.90,11.03] (p = 0.35)	-	-
Outpatient Physical, Occupational, or Speech and Language Pathology Therapy	-0.84 [-1.79,0.12] (p = 0.09)	-12.34[-36.04,11.36] (p = 0.31)	-	-
Laboratory, Pathology, and Other Tests	-0.07 [-0.21,0.07] (p = 0.32)	0.55 [-2.84,3.94] (p = 0.75)	0.03 [-0.04,0.11] (p = 0.38)	-0.97 [-2.74,0.80] (p = 0.28)
Imaging Services	0.50 [0.28,0.72] (p < 0.01)	9.25 [3.67,14.83] (p < 0.01)	0.30 [0.17,0.42] (p < 0.01)	5.18 [2.24,8.11] (p < 0.01)
Durable Medical Equipment and Supplies	0.09 [0.04,0.14] (p < 0.01)	-2.42 [-3.62,-1.21] (p < 0.01)	0.07 [0.04,0.09] (p < 0.01)	-1.98 [-2.62,-1.34] (p < 0.01)
Inpatient Hospital Trigger	0.03 [0.03,0.03]	-0.09 [-0.13,-0.05]	0.04 [0.04,0.04]	-0.05 [-0.07,-0.03]

	(p < 0.01)	(p < 0.01)	(p < 0.01)	(p < 0.01)
Physician Services During Hospitalization Trigger	0.04 [0.03,0.04] (p < 0.01)	0.11 [0.00,0.22] (p = 0.06)	0.05 [0.05,0.05] (p < 0.01)	-0.14 [-0.20,-0.08] (p < 0.01)
Anesthesia Services	0.40 [0.17,0.64] (p < 0.01)	-0.67 [-6.50,5.15] (p = 0.82)	0.46 [0.34,0.58] (p < 0.01)	6.00 [3.18,8.82] (p < 0.01)
Chemotherapy and Other Part B-Covered Drugs	0.15 [0.06,0.24] (p < 0.01)	-1.15 [-3.29,1.00] (p = 0.29)	0.05 [0.01,0.08] (p = 0.01)	0.94 [0.16,1.73] (p = 0.02)
Dialysis	0.00 [-0.11,0.11] (p = 1.00)	-0.19 [-2.82,2.43] (p = 0.89)	0.13 [0.04,0.22] (p < 0.01)	4.54 [2.44,6.64] (p < 0.01)
All Other Services Not Otherwise Classified	-0.14 [-0.64,0.35] (p = 0.57)	-8.35 [-20.67,3.97] (p = 0.18)	-0.22 [-0.51,0.07] (p = 0.14)	-1.65 [-8.29,4.98] (p = 0.63)

3.3.4 Interpretation

Overall, the results demonstrate that the cost measure is reflective of both the cost directly related to treatment choices, as well as cost of adverse outcomes as a result of care (Table 6).

Model 1 shows that the cost of adverse events is associated with a worse measure score. The costs of outpatient (OP) evaluation and management (E/M) and imaging services are associated with worse measure scores (Model 1) and higher costs of adverse events (Model 2), which suggests that avoidance of adverse events may also reduce spending related to these services and improve measure performance. Meanwhile, other services, such as durable medical equipment and supplies (DME), Part B medications, and dialysis, are associated with worse measure scores, but they do not appear to be associated with adverse events, which suggests that overuse of these services may negatively impact measure scores. These results suggest that the measure is capturing what it aims to measure.

The following service categories were not identified as being associated with the Respiratory Infection Hospitalization measure score or adverse events at the TIN level: Ambulatory/Minor Procedures and Outpatient Physical, Occupational, or Speech and Language Pathology Therapy. Similarly, Laboratory, Pathology, and Other Tests, and All Other Services Not Otherwise Classified were not identified as associated with the measure score or adverse events at the TIN-NPI level.

3.4 Exclusions Analysis

3.4.1 Method of Testing Exclusions

Exclusions are used in the Respiratory Infection Hospitalization measure to ensure a comparable patient population within the scope of the measure's focus on patients who receive inpatient treatment for a respiratory infection and that episodes provide meaningful information to attributed clinicians. Exclusions are also used as part of data processing so that sufficient data are available to accurately determine episode spending and calculate risk adjustment for each episode.

For the exclusions analysis discussed in this section, we focused on exclusion criteria intended to ensure a comparable patient population.

- Episodes where patient death date occurred before the episode end date
 - These episodes were excluded as they may not accurately reflect a clinician's performance as the truncated episode window does not capture the full length of care intended by the measure.
- Episodes with patients with pleurisy diagnosis, pleural conditions, pleural plaque, chest trauma, chest wall myopathy, epidemic myalgia, fibrothorax, influenza due to avian flu, adverse effects of glucocorticoids, hospitalizations for certain non-pneumonia infection/reaction diagnoses.
 - These episodes may not accurately reflect a clinician's performance and were excluded as these cases may substantially deviate from the projected cost for a given patient risk profile.
- Episodes with patients discharged against medical advice.
 - These episodes may not accurately reflect a clinician's performance and were excluded as these cases may be beyond a clinician's influence.
- Episodes with patients with overlapping IP admission days or who are treated at non-acute hospital, psychiatric facilities
 - These episodes were excluded as they may be influenced by exceptional payments that substantially deviate from the projected cost.

Given the rationales for these exclusions, we expect these excluded episodes to have a different profile than the included episodes, such as a higher mean cost, or a different distribution of costs (e.g., a long tail of high-cost episodes). For each exclusion, we examined the number of episodes and beneficiaries affected, as well as the distributions of observed cost. We then compared the cost characteristics of the excluded episodes to those of episodes included in the measure calculation to assess the distinctness between the two patient cohorts. A full list of the exclusions used for the Respiratory Infection Hospitalization measure is provided in the Measure Codes List available on the [QPP Cost Measure Information Page](#).³¹

3.4.2 Statistical Results from Testing Exclusions

Table 7 below presents descriptive statistics of all episodes meeting the measure's triggering logic, excluded episodes, and final reportable episodes at both TIN and TIN-NPI levels. These exclusion criteria ensure that the reportable episode populations are more homogenous and comparable than all episodes meeting triggering logic.

³¹CMS, QPP Cost Measure Information Page, <https://www.cms.gov/medicare/quality/value-based-programs/cost-measures>.

Table 7: Cost Statistics for Measure Exclusions

Exclusion	Episodes		Mean	Observed Cost				
	#	% of All Episodes Meeting Triggering Logic		Percentile				
				10 th	25 th	50 th	75 th	90 th
All Episodes Meeting Triggering Logic	469,993	100%	\$15,796	\$7,396	\$9,833	\$13,168	\$18,199	\$28,078
Beneficiary Death in Episode	61,895	13.17%	\$16,488	\$8,997	\$11,520	\$13,967	\$18,209	\$26,472
Outlier	7,060	1.50%	\$34,120	\$4,868	\$6,480	\$23,432	\$55,701	\$68,997
No Attributed TIN-NPI	34,008	-	\$12,888	\$5,679	\$8,479	\$11,979	\$12,889	\$22,401
Not an IPPS Acute Hospital or Psychiatric Facility	37,588	8.00%	\$18,232	\$6,588	\$8,949	\$13,096	\$22,783	\$37,250
Overlapping IP Admission Days	915	0.19%	\$16,915	\$5,182	\$6,460	\$14,323	\$20,093	\$32,208
TIN does not Meet Testing Volume Threshold	51,948	11.05%	\$16,541	\$6,649	\$9,117	\$12,919	\$19,827	\$30,558
TIN-NPI does not Meet Testing Volume Threshold	247,286	52.61%	\$15,447	\$7,045	\$9,506	\$12,913	\$17,312	\$27,630
Principal Diagnosis of Abscess	202	0.04%	\$16,638	\$7,353	\$9,373	\$13,040	\$20,437	\$27,208
Adverse effects of glucocorticoids and synthetic analogues (T380X5)	6,434	1.37%	\$17,949	\$8,353	\$10,391	\$13,833	\$21,581	\$31,500
Pleural Plaque with Presence of Asbestos	15	0.00%	\$11,657	\$6,758	\$7,411	\$9,804	\$11,273	\$26,548
Principal Diagnosis of Invasive Pulmonary Aspergillosis	63	0.01%	\$22,526	\$10,831	\$13,711	\$15,814	\$26,126	\$42,765
Principal Diagnosis of Pulmonary Candidiasis	118	0.03%	\$20,485	\$10,619	\$13,437	\$15,820	\$22,185	\$38,546
Chest Trauma	5,286	1.12%	\$16,847	\$7,569	\$10,070	\$13,594	\$21,558	\$29,791
Principal Diagnosis of Coccidioidomycosis	143	0.03%	\$14,918	\$8,971	\$11,042	\$13,741	\$16,932	\$23,391
Principal Diagnosis of Cystic Fibrosis with Pulmonary Manifestations	511	0.11%	\$15,029	\$8,349	\$9,470	\$13,130	\$14,987	\$22,057
Principal Diagnosis of Gangrene and Necrosis of Lung	352	0.07%	\$17,567	\$9,055	\$12,401	\$14,660	\$19,267	\$30,301
Principal Diagnosis of Histoplasmosis Capsulati	53	0.01%	\$21,494	\$9,289	\$13,336	\$16,509	\$27,881	\$42,430
Influenza due to Avian Flu	390	0.08%	\$12,096	\$5,963	\$6,908	\$9,661	\$12,947	\$22,728
Principal Diagnosis of Influenza Unspecified	6,224	1.32%	\$12,738	\$6,444	\$9,042	\$10,074	\$13,180	\$24,012

Exclusion	Episodes		Mean	Observed Cost				
	#	% of All Episodes Meeting Triggering Logic		Percentile				
				10 th	25 th	50 th	75 th	90 th
Discharged Against Medical Advice	3,924	0.83%	\$13,304	\$7,227	\$9,467	\$12,600	\$13,732	\$18,123
Principal Diagnosis of Legionnaires' Disease	325	0.07%	\$15,191	\$8,392	\$9,539	\$13,574	\$16,477	\$25,083
Principal Diagnosis of Mediastinitis	94	0.02%	\$12,399	\$6,606	\$8,327	\$11,789	\$14,898	\$18,110
Principal Diagnosis of Mycobacterial Infection	540	0.11%	\$15,520	\$8,667	\$9,762	\$13,486	\$16,969	\$27,429
Chest Wall Myopathy	1,041	0.22%	\$15,817	\$7,839	\$9,999	\$13,445	\$17,822	\$27,416
Principal Diagnosis of Nocardiosis	64	0.01%	\$21,965	\$9,195	\$12,561	\$15,852	\$29,744	\$37,214
Nonspecific reaction to skin test or cell mediated immunity measurement	17	0.00%	\$13,032	\$6,349	\$8,475	\$11,308	\$14,693	\$26,959
Pleurisy	148	0.03%	\$7,950	\$5,158	\$6,013	\$6,525	\$8,726	\$10,761
Principal Diagnosis of Pneumocystosis	311	0.07%	\$19,939	\$10,194	\$13,926	\$15,444	\$21,474	\$35,507
Principal Diagnosis of Cytomegaloviral Pneumonitis	31	0.01%	\$20,457	\$8,772	\$11,171	\$16,054	\$24,778	\$33,083
Principal Diagnosis of Pyothorax	1,798	0.38%	\$20,138	\$9,592	\$13,541	\$15,617	\$23,760	\$38,011
Principal Diagnosis of Tuberculosis	187	0.04%	\$17,311	\$8,848	\$10,298	\$14,575	\$19,256	\$27,660
Reportable Episodes (if all clinicians reported as TIN at the Testing Volume Threshold)	318,663	67.80%	\$14,995	\$7,470	\$9,756	\$13,017	\$17,232	\$27,071
Reportable Episodes (if all clinicians reported as TIN-NPI at the Testing Volume Threshold)	170,799	36.34%	\$15,407	\$7,785	\$9,997	\$13,221	\$18,180	\$27,625

3.4.3 Interpretation

The statistical results show that applying the above exclusion criteria decreases the cost of all episodes meeting trigger logic, from the observed mean of \$15,796 to \$14,995 at the TIN-level and \$15,407 at the TIN-NPI level, closer to the expected costs for respiratory infection hospitalization. This supports the exclusion of these episodes to ensure a comparable patient cohort that will yield a clinically coherent measure and meaningful information to attributed clinicians. Further discussion of the results for exclusions applied based on the clinical validity of the study population are provided below.

Most of the excluded episodes regard small populations and make up less than 1% of all episodes meeting the trigger logic. However, most also had a higher mean observed cost than all episodes meeting the trigger logic. In particular, the largest mean observed costs came from episodes that were outliers, pneumocystosis, cytomegaloviral pneumonitis, pyothorax, nocardiosis, histoplasmosis capsulati, pulmonary, candidiasis, or invasive pulmonary aspergillosis.

Episodes classified as outlier cases were excluded because they deviate substantially from the projected cost for a given patient risk profile, as seen by their high mean observed cost of \$34,120 and their wide cost variability. At the 10th percentile the observed cost is \$4,868 and at the 90th percentile the observed cost is \$68,997.

Episodes where a beneficiary had pleurisy, pneumocystosis, cytomegaloviral pneumonitis, pyothorax, nocardiosis, histoplasmosis capsulati, pulmonary, candidiasis, or invasive pulmonary aspergillosis were excluded because their care may substantially deviate from an average patient with respiratory infection. Most of these episodes had higher mean observed costs and large cost distributions. For example, episodes with patients that had invasive pulmonary aspergillosis had a mean observed cost of \$22,526 and ranged from \$10,831 in the 10th percentile to \$42,765 in the 90th percentile. However, the mean observed cost for episodes with patients that had pleurisy was \$7,950, only ranging to \$10,761 in the 90th percentile. This is much lower than for all episodes meeting trigger logic.

Only the observed cost is shown, which has not been risk adjusted. The differences in cost may appear much smaller after risk adjustment.

3.5 Risk Adjustment or Stratification

3.5.1 Method of Controlling for Differences

Differences in case mix are controlled for using a statistical risk model with 113 risk factors and stratification by 2 risk categories.

The risk adjustment model for the Respiratory Infection Hospitalization measure adjusts for comorbidities based on the CMS Hierarchical Condition Category (HCC) model, count of HCCs, end-stage renal disease (ESRD) status, disability status, number and types of clinician specialties from which the patient has received care, recent use of institutional long-term care, and age.

The model also includes measure-specific factors:

- Asthma
- Acid-base disorders
- COVID-19
- Pleural effusion/thoracentesis
- Dementia
- Limited mobility
- Recent use of long-term assisted care within 30 days
- Recent all-cause admission in prior 120 days
- Prior oxygen use/respiratory failure

A separate linear regression is run for each sub-group to ensure fair comparison:

- Respiratory Infections and Inflammations

- Simple Pneumonia and Pleurisy

The episode's scaled (i.e., annualized) observed costs are winsorized at the 98th percentile prior to the regression for each model to handle extreme observations. Full details of the risk adjustment model are in the Measure Codes List File available on the [QPP Cost Measure Information page](#).³²

3.5.2 Conceptual, Clinical, and Statistical Methods

We selected the CMS-HCC model based on previous studies evaluating its appropriateness for use in risk adjusting Medicare claims data. This model was developed specifically for use in the Medicare population, meaning that it accounts for conditions found in the Medicare population. In addition, the CMS-HCC model is routinely updated for changes in coding practices (e.g., the transition from ICD-9 to ICD-10 codes). Because the CMS-HCC model has already been extensively tested, we focus our testing on the adaptation of the CMS-HCC model to the Respiratory Infection Hospitalization measure's patient population.

The workgroup provided input on measure-specific risk adjusters after reviewing empirical analyses on subpopulations of interest to assess whether and if so, how, particular factors should be accounted for in the model. These could include patient characteristics, factors outside of the reasonable influence of the clinician, or any other factors that would help prevent unintended consequences. These additional risk adjusters are listed in the section above.

As previously noted, the risk adjustment model is run on episodes stratified into episode sub-groups, which may qualify as "ordering" of risk factors. Episode sub-groups were also determined based on the workgroup's input, with the goal of ensuring clinical comparability among episodes so that the cost measure fairly compares clinicians with similar patient case-mix.

3.5.3 Conceptual Model of Impact of Social Risks

Figure 3 shows the conceptual model that outlines how SRFs can influence the measure score, which is informed by published external research and Acumen's data analysis.^{23,33,34,35,36} The conceptual model outlines risk factors that are either known by the literature or informed by the Clinical Expert Workgroup to be within or outside the influence of the attributed clinician. Risk factors, including SRFs, can influence the treatment choices and impact the size of the effect of treatment choices on mitigating the risk and cost of adverse outcomes.

A systematic approach then guides the decision of which factors to include in the risk adjustment model:

1. First, we reviewed the literature to gather known risk factors and drivers of resource use. These factors are usually diagnoses. Therefore, the first set of risk adjusters are commonly the HCCs.

³²CMS, QPP Cost Measure Information Page, <https://www.cms.gov/medicare/quality/value-based-programs/cost-measures>.

³³Assistant Secretary of Health and Human Services for Planning and Evaluation. Report to Congress: Social Risk Factors and Performance Under Medicare's Value-Based Purchasing Programs. Washington, D.C. December 2016.

³⁴Chen LM, Epstein AM, Orav EJ, Filice CE, Samson LW, Joynt Maddox KE. Association of Practice-Level Social and Medical Risk With Performance in the Medicare Physician Value-Based Payment Modifier Program. JAMA. 2017;318(5):453-461

³⁵Medicare Payment Advisory Commission. Beneficiaries Dually Eligible for Medicare and Medicaid. 2018; <https://www.macpac.gov/publication/data-book-beneficiaries-dually-eligible-for-medicare-and-medicaid-3/>.

³⁶Office of the Assistant Secretary for Planning and Evaluation, U.S. Department of Health & Human Services. Second Report to Congress on Social Risk Factors and Performance in Medicare's Value-Based Purchasing Program. 2020. <https://aspe.hhs.gov/social-risk-factors-and-medicare-value-based-purchasing-programs>

2. Then, we consulted our clinical expert panels on additional factors that are known to be associated with resource use. Together with our clinical expert panel, we reviewed the stratified results on episode cost across many patient characteristics. We arrived at the final list of risk adjusters based on those discussions and consensus among the clinical experts.
3. During our testing phases, we also follow a structured and systematic approach to deciding whether SRFs should be adjusted for, further described in Section 3.5.5.

3.5.4 Statistical Results

The literature has extensively tested using the HCC model for Medicare claims data. Although the variables in the HCC model were selected to predict annual cost, CMS has also used this risk adjustment model in several other settings (e.g., Accountable Care Organizations, previous physician Quality and Resource Use Report programs, and other administrative claims-based measures such as the Knee Arthroplasty episode-based cost measure, Total Per Capita Cost (TPCC) cost measure, Medicare Spending Per Beneficiary (MSPB)-PAC cost measure and MSPB-Hospital cost measure). Recalling that the risk model relies on the existing CMS-HCC model, testing results for factors included in the CMS-HCC V24 model can be found in the Evaluation of the CMS-HCC Risk-Adjustment Model report³⁷ and the Report to Congress: Risk Adjustment in Medicare Advantage³⁸. For measure-specific factors not included in the CMS-HCC model, we sought expert clinician input through the workgroup, which provided recommendations on additional risk adjusters and sub-groups.

3.5.5 Analyses and Interpretation in Selection of Social Risk Factors

To determine whether it is appropriate to risk adjust for SRFs, the following criteria are considered:

- (i) whether there is an association between social risk and performance by examining the coefficient of patient-level dual status when added into the risk model,
- (ii) whether the observed association is most influenced by patient-level factors or clinician-level factors by examining the stability of the patient-level dual status coefficient after adding clinician's dual share variable, as well as including clinician's fixed effects,
- (iii) whether patient's need or complexity rather than poor quality is driving the observed performance differences by examining the differences in performance on dual patients versus non-dual patients and if there are many clinicians who are able to perform similarly or better on their dual patients than their non-dual patients, and
- (iv) the impact of risk adjusting for SRFs by examining the performance shift of clinicians compared to a risk adjustment model that does not risk adjust for SRFs.

³⁷Pope, Gregory C., John Kautter, et al., "Evaluation of the CMS-HCC Risk-Adjustment Model: Final Report." RTI International: March 2011.

³⁸CMS, "Report to Congress: Risk Adjustment in Medicare Advantage," <https://www.cms.gov/Medicare/Health-Plans/MedicareAdvtgSpecRateStats/Downloads/RTC-Dec2018.pdf>.

Table 8: Coefficient of Patient-level Dual Status under Different Models

Reporting Level	Subgroup Risk Model	% of All Episodes	Coefficient of Patient-level Dual Status		
			Base Model + Patient-level Dual Status	Base Model + Patient-level Dual Status + Clinician's Dual Share	Base Model + Patient-level Dual Status + Clinician's Fixed Effect
TIN	Respiratory Infections and Inflammations	29.05%	\$843 (p < 0.0001)	\$422 (p < 0.0001)	\$360 (p < 0.0001)
TIN	Simple Pneumonia and Pleurisy	26.63%	\$551 (p < 0.0001)	\$336 (p < 0.0001)	\$346 (p < 0.0001)
TIN-NPI	Respiratory Infections and Inflammations	28.42%	\$691 (p < 0.0001)	\$336 (p < 0.0001)	\$277 (p < 0.0001)
TIN-NPI	Simple Pneumonia and Pleurisy	25.70%	\$478 (p < 0.0001)	\$320 (p < 0.0001)	\$410 (p < 0.0001)

Table 9: Mean Ratio of Episode Observed Cost to Expected Cost (O/E) Stratified by Clinician's Dual Share and Patient's Dual Status

Dual Share	TIN			TIN-NPI		
	All Episodes	Dual Episodes	Non-Dual Episodes	All Episodes	Dual Episodes	Non-Dual Episodes
(ALL)	1.01	1.03	1.01	0.99	1.01	0.99
0%-20%	1.00	1.03	1.00	0.98	1.03	0.97
21%-40%	1.01	1.03	1.00	0.98	1.01	0.98
41%-60%	1.00	1.01	0.99	0.99	1.01	0.99
61%-80%	1.01	1.02	1.00	1.00	1.01	0.99
81%-100%	1.04	1.05	1.05	1.01	1.01	1.01

Table 10. Proportions of Clinicians Who Perform Significantly Worst, Equally Well, or Significantly Better on Their Dual Episodes than Non-Dual Episodes

Reporting Level	Significantly Worse	Equally Well	Significantly Better
TIN	5.47%	92.92%	1.61%
TIN-NPI	4.53%	94.43%	1.03%

Table 11. Clinicians' Performance Shift after Adding a Dual Status Risk Adjustor

TIN or TIN-NPI	Proportion of Clinicians Affected at Various Levels of Performance Shift	
	Ranking Shift by 1% or more	Ranking Shift by 5% or more
TIN	71.39%	4.23%

TIN or TIN-NPI	Proportion of Clinicians Affected at Various Levels of Performance Shift	
	Ranking Shift by 1% or more	Ranking Shift by 5% or more
TIN-NPI	60.21%	2.04%

There's a statistically significant association between a patient's dual status and episode cost for episodes in both sub-groups (Table 8). This association is stable in both sub-groups, as they maintain statistical significance even after adding variables to account for clinician-level factors. Additionally, the coefficients decrease as clinician-level factors are added. These results suggest that the patient-level factors are more influential than clinician-level factors for both sub-groups. However, both dual and non-dual episodes remain relatively stable as clinician dual share increases (Table 9). Also, Table 10 shows that many clinicians perform equally well for dual and non-dual episodes and some even perform significantly better on dual episodes. Lastly, risk adjusting for dual status appears to change measure performance for many clinicians, but few clinician's ranks shift by 5% or more (Table 11). These results suggest that clinicians are able to mitigate many effects of SRFs.

3.5.6 Method for Statistical Model or Stratification Development

To analyze the validity of current risk adjustment model, we examined two criteria: discrimination and calibration.

- 1) Discrimination is a statistical criterion that evaluates the measure's ability to distinguish high-cost episodes from low-cost episodes, or the ability to explain the variance in cost of individual episodes. The amount of variance explained is estimated by the R-squared metric with the range between 0 and 1. These results are provided in Section 3.5.7.
- 2) Calibration evaluates the consistency of the measure in estimating episode cost across the full range of resource use patterns in the population. Calibration is estimated by the average predictive ratios across groups within the population, specifically groups are partitioned by deciles of expected episode cost. A well-calibrated measure should have predictive ratios close to 1.0 across all deciles. These are discussed in Sections 3.5.8 and 3.5.9.

3.5.7 Statistical Risk Model Discrimination Statistics

The overall R-squared for the Respiratory Infection Hospitalization cost measure, calculated by dividing explained sum of squares by total sum of squares is 0.23. The adjusted R-squared is 0.22. More information on discrimination testing for the CMS-HCC model can be found at Pope et al. 2011.³⁹

3.5.8 Statistical Risk Model Calibration Statistics

The predictive ratio is calculated using the formula of average expected cost / average observed cost for all episodes in each decile.

³⁹Pope, Gregory C., John Kautter, et al., "Evaluation of the CMS-HCC Risk-Adjustment Model: Final Report." RTI International: March 2011.

3.5.9 Statistical Risk Model Calibration – Risk Decile

Analysis of predictive ratios by risk decile for the measure shows minimal variation among risk deciles, as predictive ratios range from 0.97 to 1.02 across all risk deciles (with an overall average of 1.00).

Table 12: Predictive Ratio by Decile of Predicted Episode Cost

Decile	Average Predictive Ratio
Decile 1	1.02
Decile 2	1.01
Decile 3	1.01
Decile 4	0.99
Decile 5	1.00
Decile 6	1.02
Decile 7	1.01
Decile 8	1.00
Decile 9	0.99
Decile 10	0.97

3.5.10 Interpretation

The R-squared values for the model, which measure the percentage of variation in results predicted by the model, are higher than the values presented in similar analyses of risk adjustment models.⁴⁰ As noted in Section 3.5.6 and 3.5.7, these results should be interpreted alongside service assignment rules, which remove clinically unrelated services.

The remaining unexplained variance is due to variation in factors that are not adjusted for by the measure, such as the clinician's performance. The objective of a cost measure is to evaluate and differentiate the performance of clinicians. Therefore, achieving high explained variance is optional because the measure should only adjust for some variations in the cost of care. In collaboration with the experts from our clinical workgroup, this measure only adjusts for factors that are deemed outside the reasonable influence of clinicians. The service assignment rules provide context for which costs are included in the measure and which are not.

Table 12 shows that the risk adjustment model is consistent, with the average predictive ratios observed to be close to 1.00 across all deciles, with the range between 0.97 and 1.02. Overall, the risk adjustment model does not over- or under-predict cost across the full range of resource use patterns in the population.

3.6 Identification of Meaningful Differences in Performance

3.6.1 Method

To identify meaningful differences in performance, this analysis first examines the distribution of the measure score to highlight the performance gap between the most and least efficient clinicians. Then, this analysis examines the rate of adverse events that may occur during an episode of care to highlight the variation in frequency and cost of those events.

⁴⁰Pope, Gregory C., John Kautter, Melvin J. Ingber, Sara Freeman, Rishi Sekar, and Cordon Newhart. "Evaluation of the CMS-HCC Risk-Adjustment Model: Final Report." RTI International: March 2011.

3.6.2 Statistical Results

Table 1 shows the distribution of the measure score at the TIN and TIN-NPI levels. There is a difference in mean score for TIN and TIN-NPI levels because each level has its own attribution rules, which resulted in slightly different populations of episodes used for measure score calculation (Table 1). However, clinicians are only compared to their peers at either the TIN or TIN-NPI level, therefore the differences in score across different levels can be ignored.

While a small percent of episodes had an assigned readmission (2.96%), the associated average observed episode cost is \$8,242 more than the average respiratory infection hospitalization episode. Similarly, the rate of episodes with inpatient (IP) rehabilitation or long-term care hospital (LTCH) services is observed to be at 1.15%, but it costs an average of \$19,140 more. Lastly, the rate of episodes with skilled nursing facility (SNF) services is high, 19.81%, costing an average of \$10,287 more.

3.6.3 Interpretation

There is substantial variation observed in the measure score in both TIN and TIN-NPI levels, indicated by the interquartile ranges, standard deviations, and coefficients of variation. The magnitude of the observed variation is in the thousands of dollars, which indicates that there are opportunities to close the gaps between the most and least efficient clinicians.

Since episodes with readmissions, SNF services, and IP rehabilitation or LTCH service have high observed costs, every percentage reduction in their rates represents substantial performance improvement for the attributed clinician or clinician group.

3.7 Missing Data Analysis and Minimizing Bias

3.7.1 Method

Since CMS uses Medicare claims data to calculate the Respiratory Infection Hospitalization measure, Acumen expects a high degree of data completeness. To further ensure that we have complete and accurate data for each patient, Acumen excludes episodes where patient date of birth information (an input to the risk adjustment model) cannot be found in the EDB, the patient does not appear in the EDB, or the patient death date occurs before the episode trigger date.

The Respiratory Infection Hospitalization measure also excludes episodes where the patient is enrolled in Medicare Part C or has a primary payer other than Medicare in the 120-day lookback period and episode window. In such situations, Medicare Parts A and B claims data may not capture the complete clinical profile for the patient needed to capture the clinical risk of the patient in risk adjustment. Furthermore, Parts A and B claims data may not capture all Medicare resource use if some portion of the patient's care is covered under Medicare Part C.

3.7.2 Missing Data Analysis

The table below presents the frequency of missing data across the categories of missing data which caused episodes to be excluded from the Respiratory Infection Hospitalization measure. Frequency is presented in terms of the number of episodes excluded due to missing data, as well as the cost profile of episodes with missing data compared to episodes included in the measure reporting.

As a note, the episode and clinician counts below reflect exclusion from the initial population of triggered episodes. After the missing data exclusions are applied, we apply additional exclusions, as outlined in section 3.4, to this overall patient cohort to narrow the population to only applicable episodes.

Table 13: Cost Statistics for Missing Data Category

Missing Data Categories	Episodes	Observed Cost					
	#	Mean	Percentile				
			10 th	25 th	50 th	75 th	90 th
Primary Payer Other than Medicare	47,326	\$14,646	\$7,048	\$9,299	\$12,489	\$15,518	\$25,754
Beneficiary Death before Trigger	924	\$10,965	\$6,441	\$9,084	\$12,185	\$12,838	\$13,111
No Continuous Enrollment in Medicare Parts A and B, and Any Enrollment in Part C	43,660	\$14,304	\$6,656	\$8,761	\$11,981	\$15,082	\$25,364

3.7.3 Interpretation

The results show that the missing data episodes don't appear to be substantially different than all episodes in the initial population in terms of cost (Table 13). Given their limited frequencies, the impact of removing these episodes on the overall measure should be minimal while ensuring that clinicians are fairly evaluated on episodes with complete data.

4.0 Feasibility

4.1 Data Elements Generated as Byproduct of Care Processes

The data elements used in this measure are pulled from Medicare claims. They can be based on information generated, collected and/or used by healthcare personnel during the provision of care (e.g., diagnoses), which are then translated into the appropriate coding system (e.g. ICD-10 diagnoses, MS-DRGs) for use in Medicare claims by either the original healthcare personnel or another individual.

4.2 Electronic Sources

All data elements are in defined fields in electronic claims.

4.3 Data Collection Strategy

4.3.1 Data Collection Strategy Difficulties

Lessons and associated modifications may be categorized into three types: data collection procedures, handling of missing data, and sampling data associated with beneficiaries who died during an episode of care.

4.3.1.1 Data Collection

Acumen receives claims data directly from the CWF maintained at the CMS Baltimore Data Center. Healthcare providers submit Medicare claims to a Medicare Administrative Contractor (MAC), which are subsequently added to the CWF. However, these claims may be denied or disputed by the MAC, leading to changes to historical CWF data. In rare circumstances, finalizing claims may take many months or even years. As such, it is not practical to wait until all claims for a given month are finalized before calculating the measure, resulting in a trade-off between efficiency (accessing the data on time) and accuracy (waiting until most claims are finalized) when determining the duration (i.e., the “claims run-out” period) after which to pull claims data. To determine the appropriate claims run-out period, Acumen has tested the delay between claim service dates and claims data finalization. Based on this analysis, Acumen uses a run-out period of three months after the end of the calendar year to collect data for development and testing purposes. If CMS adopts this measure for use in a program, calculation and reporting would align with the program’s reporting practices.

4.3.1.2 Missing Data

This measure requires complete beneficiary information, therefore, a small number of episodes with missing data are excluded to ensure data completeness and accurate comparability across episodes. For example, episodes where the beneficiary was not enrolled in Medicare Parts A and B for the 120 days before the episode start date are excluded from this measure. Excluding these episodes enables the risk adjustment model to accurately adjust for the beneficiary’s comorbidities using data from the previous 120 days of Medicare claims. Additionally, the risk adjustment model includes a categorical variable for beneficiary age bracket, so episodes for which the beneficiary’s date of birth cannot be located are excluded from the measure.

4.3.1.3 Sampling

During measure testing, Acumen noted that episodes in which the beneficiary died before the episode end date exhibited different cost distributions than other episodes. As such, this measure excludes episodes to avoid negatively impacting clinician scores.

5.0 Usability and Use

5.1 Use

5.1.1 Current and Planned Use

A previous version of this measure is currently in use in MIPS. However, this measure has been revised as part of the comprehensive re-evaluation process specifically for potential use in the cost performance category of MIPS to assess clinicians reporting as individuals or groups under a contract with CMS.

For CMS to approve this measure for use in MIPS, it must be reviewed by the Pre-Rulemaking Measure Review and Measure Set Review process (PRMR-MSR; formerly referred to as the Measure Application Partnership [MAP]) and then undergo the notice-and-rulemaking process. Given these next steps, the earliest the measure could be used in MIPS is CY 2025. If in use, CMS can then determine whether to publicly report the cost measure.

5.1.2 Feedback on the Measure by Those being Measured or Others

Throughout the Respiratory Infection Hospitalization measure re-evaluation, we used an iterative and extensive process to gather feedback on the measure and its results to ensure that it can be used appropriately in the MIPS program by clinicians and clinician groups who practice in this clinical area. This process also seeks to ensure that the measured entities can understand and interpret their performance results to help support decision-making. A couple of the main ways we gathered input was through reoccurring Clinician Expert Workgroup meetings, which incorporated feedback from the patient and caregiver perspective, empirical data, and discussion between clinician experts who recommend measure specifications, and through public comment periods for the measures.

5.1.2.1 Technical Assistance Provided During Development or Implementation

Clinician Expert Workgroup Meetings

For each Clinician Expert Workgroup meeting, Acumen provided empirical data (e.g., analyses on potentially relevant revisions for the measure) to inform the Clinician Expert Workgroup members' recommendations. These analyses were conducted using all administrative claims data for Medicare Parts A and B. This data was shared with Workgroup members to help inform their feedback on the measure specifications throughout its re-evaluation to ensure that the measure is appropriately assessing costs for these clinicians.

Public Comment Period

Additionally, Acumen and CMS provided two public comment periods to gather feedback the measure's re-evaluation. The first public comment period was held from February 25, 2022 to May 28, 2022, to identify which measures in use in MIPS require re-evaluation and potential revisions to those measures. A second public comment period was held in February 2023, where interested parties were invited to submit feedback via an online survey on the potential revision before consideration of their potential use in the cost performance category of the MIPS. During this feedback period, interested parties had the opportunity to view (i) measure specifications documentation, (ii) measure testing forms, (iii) clinician expert workgroup meeting summaries, and (vi) summaries of previous Wave 1 measure feedback.

5.1.2.2 Technical Assistance with Results

Clinician Expert Workgroup Meetings

Acumen provided data before or during each of the Clinician Expert Workgroup Meetings: The Comprehensive Reevaluation Webinar, and Post-Feedback Refinement Webinar. During the

meetings, Acumen would guide Workgroup members through these analyses, providing clinical and programmatic context when needed. Using this iterative process, the Workgroup members discussed the testing results in depth during each meeting and allowed the data to inform their recommendations for measure specifications. The goal was to ensure that the measure appropriately assessed clinicians' cost of care within their reasonable influence without creating potential unintended consequences so that it could be usable in the MIPS program.

Public Comment Periods

During the February 2023 public comment period, interested parties provided feedback on the appropriateness of the measures and the usability of the data. The public comments were summarized and considered the Clinician Expert Workgroup when recommending further refinements to the measures.

Education and Outreach

Acumen directly conducted outreach via email to tens of thousands of interested parties using a contact list developed through previous public engagement efforts, as well as CMS and Quality Payment Program (QPP) listservs. Acumen also contacted specialty societies that may have interest in these measures due to the types of clinicians that they represent.

Acumen worked closely with QPP Service Center to respond to stakeholder inquiries during the public comment period and continued to answer questions after the period ended.

5.1.2.3 Feedback on Measure Performance and Implementation

Clinician Expert Workgroup Meetings

Feedback from the Workgroup members were recorded throughout the meeting. More formal feedback was gathered using polls, typically requesting for votes on certain specifications or appropriateness of the measure. These polls were conducted following each meeting and on an ad hoc basis, as needed.

Public Comment Periods

For the 2022 public comment period, Acumen received 20 comments and for the 2023 public comment period, Acumen received 18 comments. These responses included comments from specialty societies representing large numbers of potentially attributed clinicians and from individuals.

Survey responses were collected via an online survey, which contained general and detailed questions on the measure specifications.

5.1.2.4 Feedback from Measured Entities and Other Entities

Public Comment Periods

The MACRA Episode-Based Cost Measures: Comprehensive Reevaluation Public Comment Summary Report presents interested parties' feedback from the initial public comment period in 2022.⁴¹ The 2023 Revised Cost Measure Feedback Period Summary Report presents stakeholder feedback gathered during the second public comment period.⁴² The measure-specific feedback was used as the basis for refinements that were made to the measures. See Section 5.1.2.5 for refinements made to the Respiratory Infection Hospitalization measure.

⁴¹ CMS, "MACRA Episode-Based Cost Measures: Comprehensive Reevaluation Public Comment Summary Report," Cost Measures Feedback Page, <https://www.cms.gov/files/document/wave-one-public-comment-summary-report.pdf>.

⁴² CMS, "2023 Revised Cost Measure Feedback Period Summary Report," Cost Measures Feedback Page, <https://www.cms.gov/files/document/2023-revised-cost-measure-feedback-period-summary-report.pdf>.

5.1.2.5 Consideration of Feedback

Public Comments

Careful consideration was given to all feedback gathered through public comment, and several updates were made to the measure based on the recommendations of commenters and the Clinician Expert Workgroup comprised of subject matter and measure-development experts. Acumen conducted analyses into potential adjustments that could be made to the measures to improve their ability to assess the intended clinician population.

After public comment periods, Acumen compiled the feedback and provided the Clinician Expert Workgroup this information, along with the empirical analyses, to inform recommendations for any refinements needed to ensure that the measure is capturing what it was intended to capture.

The changes to the Respiratory Infection Hospitalization measure made through re-evaluation include:

- Expand the patient cohort to include beneficiaries hospitalized for pneumonia and related respiratory infections not otherwise captured under the measure due to recent changes in coding guidance

5.2 Usability

5.2.1 Improvement

The version of the measure has not yet been implemented, and as such has not had influence over performance. Our testing suggests that there is a sufficiently large difference in measure scores among clinicians to meaningfully determine a difference in performance. The potential for this measure to distinguish between good and poor performance is promising in its ability to encourage improvement in cost efficient care.

5.2.2 Unexpected Findings

There were no unexpected findings during the development and testing of this measure. This version of the measure has not been implemented at this time, so we do not have data that confirms unexpected findings related to its implementation.

However, Acumen did consider potential unintended consequences of having a cost measure for this clinical area (e.g., potential stinting in care to receive a better cost score). For example, the empiric validity data previously presented in section 3.3 demonstrates that many of the included services are not associated with the costs of adverse events, suggesting that cost improvement can be achieved without increasing occurrence of adverse events.

Additionally, CMS monitors measures that are in use and has multiple processes in place to allow for changes to a measure if appropriate. These include i) annual maintenance for non-substantial changes and upkeep, ii) ad hoc maintenance if a specific issue occurs or a large change in clinical guidance takes place, and iii) measure reevaluation every three years where the suitability of a measure's specifications is comprehensively reassessed. If in the event the measure did have any unexpected findings, it would be identified and resolved through one of these methods.

5.2.3 Unexpected Benefits

Since this version of the measure has not been implemented at this time, there are no testing results that identify unexpected benefits. However, many clinicians can only be assessed by the MSPB Clinician and TPCC measures in the cost performance category currently. This measure would provide a more tailored assessment of the care they have influence over, which many

clinicians may prefer to be measured by compared to the population-based cost measures like MSPB Clinician or TPCC.

6.0 Related and Competing Measures

6.1 Relation to Other Measures

There are no competing measures with this measure. However, the following measures have been identified as potentially related.

Table 14. Quality Measures Potentially Relevant for the Respiratory Infection Hospitalization Episode Group

Measure Title	Measure ID	Measure Description	Measure Type
Appropriate Treatment for Upper Respiratory Infection (URI)	065	Percentage of episodes for patients 3 months of age and older with a diagnosis of upper respiratory infection (URI) that did not result in an antibiotic order	Process
Hospital-Wide, 30-Day, All-Cause Unplanned Readmission (HWR) Rate for the Merit-Based Incentive Payment System (MIPS) Groups	479	This measure is a re-specified version of the measure, "Risk-adjusted readmission rate (RARR) of unplanned readmission within 30 days of hospital discharge for any condition" (NQF 1789), which was developed for patients 65 years and older using Medicare claims. This re-specified measure attributes outcomes to MIPS participating clinician groups and assesses each group's readmission rate. The measure comprises a single summary score, derived from the results of five models, one for each of the following specialty cohorts (groups of discharge condition categories or procedure categories): medicine, surgery/gynecology, cardio-respiratory, cardiovascular, and neurology.	Outcome
Documentation of Current Medications in the Medical Record	130	Percentage of visits for patients aged 18 years and older for which the eligible clinician attests to documenting a list of current medications using all immediate resources available on the date of the encounter.	Process
Advance Care Planning	047	Percentage of patients aged 65 years and older who have an advance care plan or surrogate decision maker documented in the medical record or documentation in the medical record that an advance care plan was discussed but the patient did not wish or was not able to name a surrogate decision maker or provide an advance care plan.	Process

The quality measures listed in Table 14 above are related to the Respiratory Infection Hospitalization measure as they may include metrics focused on similar patient cohorts or clinically related to the care provided for the episode group.

6.2 Harmonization

During the measure's development, the Clinician Expert Workgroup specifically considered how to align relevant cost and quality measures (e.g., episode window length). This cost measure aligns with the Patient-Focused Episode of Care goal of CMS's Meaningful Measures initiative,

and the domain of Efficiency and Cost Reduction. Through this measure, we aim to improve care by optimizing health outcomes and resource use associated with managing care during each episode of this acute inpatient medical condition. The development of episode groups for resource use analysis is also required by section 101(f) of MACRA.

6.3 Competing Measures

There are no measures that conceptually address both the same measure focus and the same target population as the Respiratory Infection Hospitalization measure.

Additional Information

Respiratory Infection Hospitalization Clinician Expert Workgroup Members:

As noted above, the following members provided detailed feedback on the measure specifications throughout its development based on public comments, clinical expertise, and empirical analyses.

Annie Perng, CRNP, CWOCN

Carolyn Fruci, MD, PhD

Heather Briggs, MD, PhD, FACP, CHCHM

Jamieson Wilcox, OTD, OTR/L

Mustafa Mark Hamed, MD, MBA, MPH, FAAFP

Measure Developer Updates and Ongoing Maintenance

The measure is not currently in use, but the earliest possible release of the measure in MIPS would be CY2025. If the measure becomes finalized for use in MIPS, it would undergo annual maintenance and a comprehensive re-evaluation every 3 years. This measure is included on the 2023 Measures Under Consideration (MUC) List and will be reviewed by PRMR in winter of 2023-2024. There are no further updates or reviews for this measure scheduled at this time.