



# Validation of claims-based algorithms to identify hospitalized COVID-19 events within the FDA Sentinel System

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- The views expressed in this presentation are those of the presenter and do not necessarily reflect those of the FDA.

# Background

- Claims databases are a rich resource of population-based information, but often lack laboratory results, the gold standard for COVID-19 diagnosis.
- There was no specific code for COVID-19 until April 1, 2020, when ICD-10-CM U07.1 was released.
- While coding guidelines were updated, diagnostic test availability and guidance evolved as well.

**Objective: Determine the ICD-10-CM code(s) that defines the most **valid** and **complete** cohort of patients hospitalized with COVID-19**

# Methods: Data

## FDA Sentinel System Data Partners

- Six data partners participated:
  - Two national claims-based insurers
  - Four integrated delivery systems (IDS)
    - Provide continuum of health care services and have associated health plan
- Selection criteria:
  - Availability of laboratory result data
  - Availability of adjudicated and unadjudicated claims

# Methods: Algorithms Evaluated

Five diagnosis code-based algorithms

	Algorithm 1	Algorithm 2	Algorithm 3	Algorithm 4	Algorithm 5
<b>U07.1</b> <i>COVID-19</i>					
<b>B97.29</b> <i>Other coronavirus as the cause of diseases classified elsewhere</i>					
<b>B34.2</b> <i>Coronavirus infection, unspecified</i>					
<b>J12.81</b> <i>Pneumonia due to SARS-associated coronavirus</i>					
<b>B97.21</b> <i>SARS-associated coronavirus as the cause of diseases classified elsewhere</i>					
<b>Pneumonia</b>					
<b>J80</b> <i>Acute Respiratory Distress Syndrome</i>					

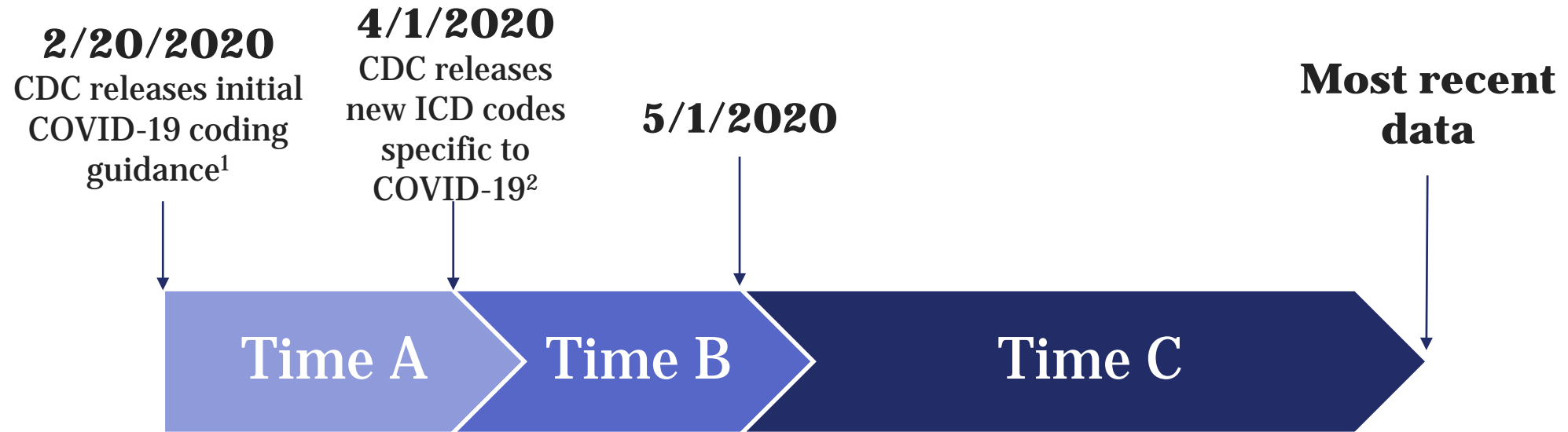
# Methods: Algorithms Evaluated

Five diagnosis code-based algorithms

	Algorithm 1	Algorithm 2	Algorithm 3	Algorithm 4	Algorithm 5
<b>U07.1</b> <i>COVID-19</i>					
<b>B97.29</b> <i>Other coronavirus as the cause of diseases classified elsewhere</i>		OR	OR	OR	OR
<b>B34.2</b> <i>Coronavirus infection, unspecified</i>			OR	OR	
<b>J12.81</b> <i>Pneumonia due to SARS-associated coronavirus</i>				OR	
<b>B97.21</b> <i>SARS-associated coronavirus as the cause of diseases classified elsewhere</i>				OR	
<b>Pneumonia</b>					OR
<b>J80</b> <i>Acute Respiratory Distress Syndrome</i>					OR

AND

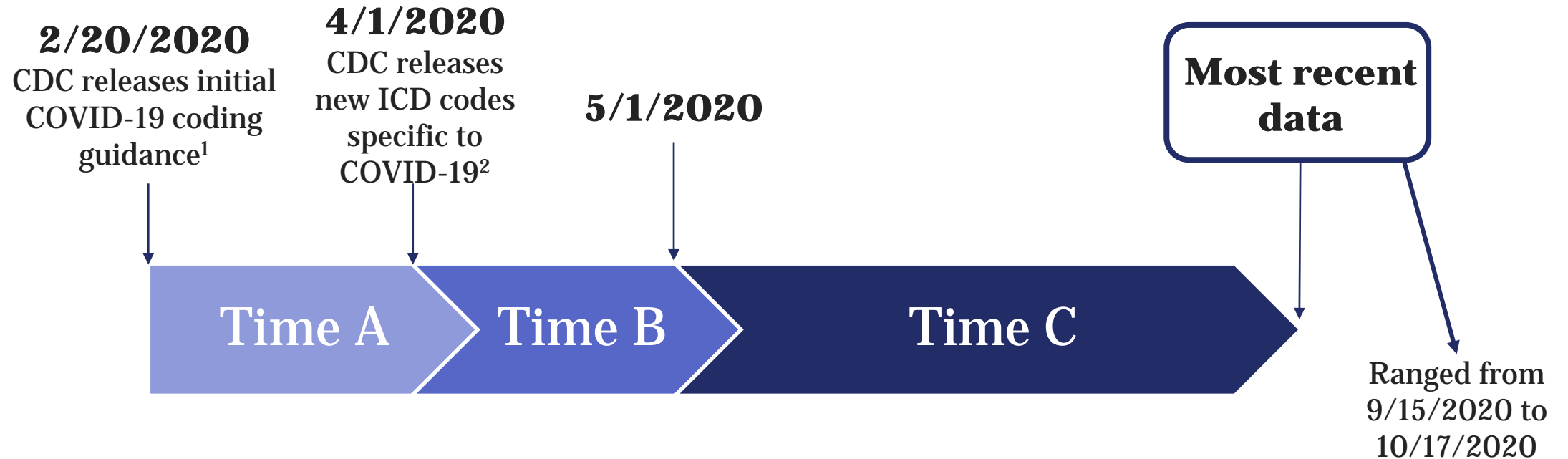
# Methods: Study Design



1. ICD-10-CM Official Coding Guidelines - Supplement. Coding encounters related to COVID-19 Coronavirus Outbreak, February 20, 2020 – March 31, 2020. Accessed November 20, 2020. <https://www.cdc.gov/nchs/data/icd/ICD-10-CM-Official-Coding-Guidance-Interim-Advice-coronavirus-feb-20-2020.pdf>

2. ICD-10-CM Official Coding and Reporting Guidelines, April 1, 2020 through September 30, 2020. Accessed November 20, 2020. <https://www.cdc.gov/nchs/data/icd/COVID-19-guidelines-final.pdf>

# Methods: Study Design



1. ICD-10-CM Official Coding Guidelines - Supplement. Coding encounters related to COVID-19 Coronavirus Outbreak, February 20, 2020 – March 31, 2020. Accessed November 20, 2020. <https://www.cdc.gov/nchs/data/icd/ICD-10-CM-Official-Coding-Guidance-Interim-Advice-coronavirus-feb-20-2020.pdf>

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# Methods: Cohort Identification

## Algorithm-based cohorts:

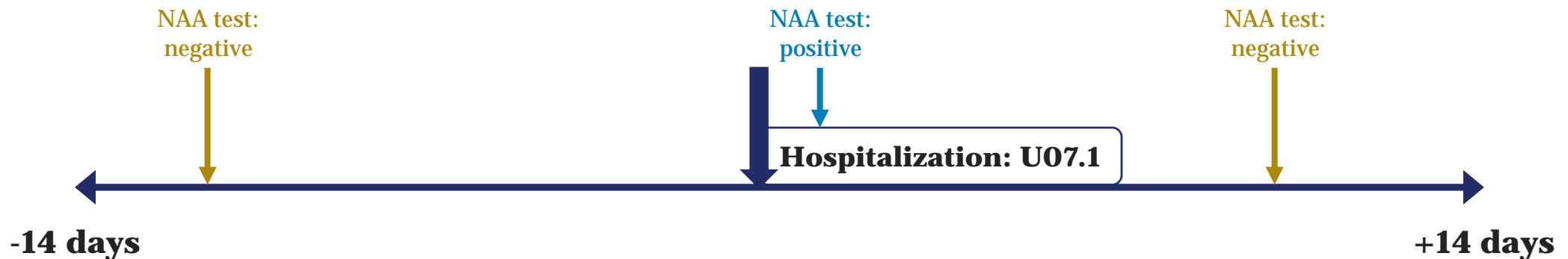
- Identify patients with algorithm-defining code(s) during a single inpatient stay
- Classify algorithm-identified patients according to observed COVID-19 nucleic acid amplification (NAA) lab results within  $\pm 14$  days of inpatient admission date:
  - **Lab-positive:** Any positive NAA test
  - **Lab-negative:** All negative NAA test(s)
  - **NAA test performed, no result to-date**
  - **No NAA test observed**



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# Methods: Cohort Identification

<b>Patients identified via Algorithm 1</b>			
Lab (+) patients	Lab (-) patients	Lab performed, results unknown	Patients with no lab
<b><i>A</i></b>	<b><i>B</i></b>	<b><i>C</i></b>	<b><i>D</i></b>

# Methods: Statistical Analysis

<b>Patients identified via Algorithm 1</b>			
Lab (+) patients	Lab (-) patients	Lab performed, results unknown	Patients with no lab
<b><i>A</i></b>	<b><i>B</i></b>	<b><i>C</i></b>	<b><i>D</i></b>

$$\text{Positive Predictive Value} = \frac{A}{A+B}$$

# Methods: Cohort Identification

## Lab-based cohorts:

- Identify patients with COVID-19 nucleic acid amplification (NAA) test within  $\pm 14$  days of **any** inpatient admission (including non-COVID-19 related hospitalizations)
- Classify patients according to NAA result:
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# Methods: Cohort Identification

Patients identified via Algorithm 1				Patients with $\geq 1$ NAA test $\pm 14$ days of hospitalization	
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<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>	<b>X</b>	<b>Y</b>

# Methods: Statistical Analysis

Patients identified via Algorithm 1				Patients with $\geq 1$ NAA test $\pm 14$ days of hospitalization	
Lab (+) patients	Lab (-) patients	Lab performed, results unknown	Patients with no lab	Lab (+) patients	Lab (-) patients
<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>	<b>X</b>	<b>Y</b>

$$\text{Sensitivity} = \frac{A}{X}$$



# Methods: Statistical Analysis

**Counts, PPV calculations, and sensitivity calculations stratified by:**

Time Period

x

- Demographics:
  - Sex
  - Age category
  - Race
  - Ethnicity
- Symptoms:
  - Pneumonia
  - Bronchitis
  - Respiratory infection
  - ARDS
  - Sepsis
  - Cough
  - Shortness of breath
  - Fever

# Results

**Algorithm 1:** U07.1  
**Algorithm 2:** U07.1 or B97.29  
**Algorithm 3:** U07.1 or B97.29 or B34.2  
**Algorithm 4:** U07.1 or B97.29 or B34.2 or J12.81 or B97.21  
**Algorithm 5:** [U07.1 or B97.29] and [pneumonia or ARDS]

## Patient counts across algorithms and over time

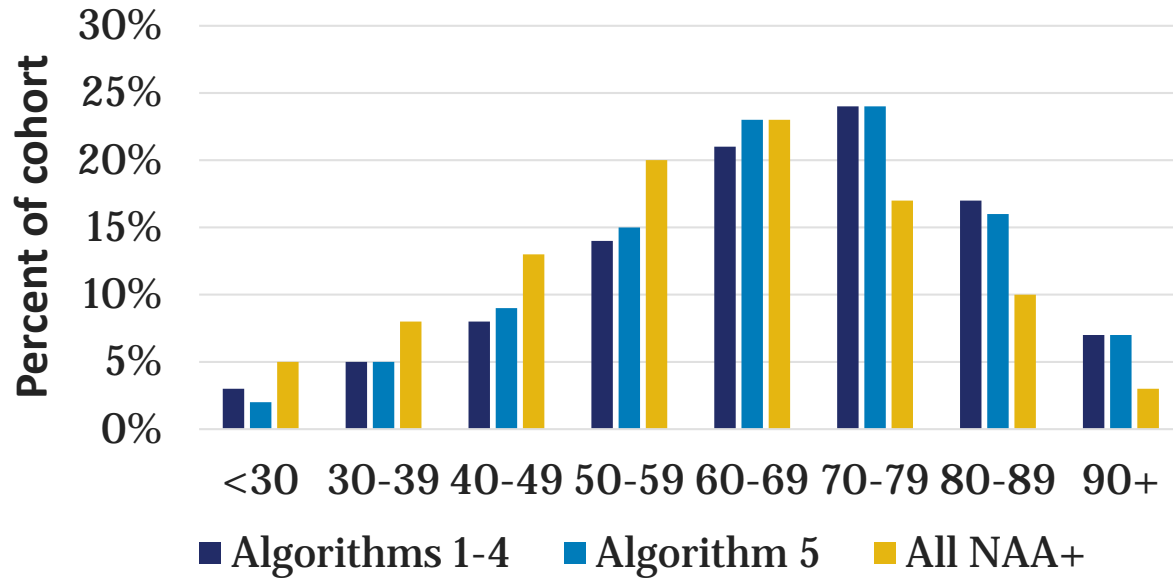
	Algorithm 1	Algorithm 2	Algorithm 3	Algorithm 4	Algorithm 5
<b>Time A</b> (2/20 - 3/31)	3,644	5,997	6,354	6,474	5,282
<b>Time B</b> (4/1 - 4/30)	12,815	12,945	13,081	13,153	10,389
<b>Time C</b> (5/1 - current)	27,238	27,424	27,588	27,666	19,135
<b>Total</b>	<b>43,697</b>	<b>46,366</b>	<b>47,023</b>	<b>47,293</b>	<b>34,806</b>

Algorithm 1 (U07.1 alone) captures almost as many patients as the broader algorithms.

# Results: Cohort demographics

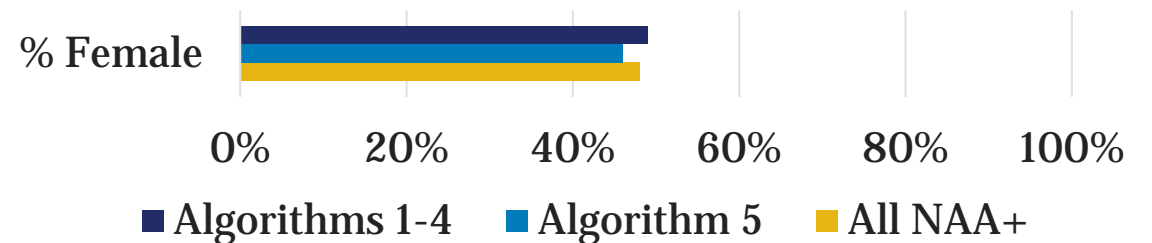
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## Age distribution



**Diagnosis code-based cohorts are slightly older than lab-based cohort**

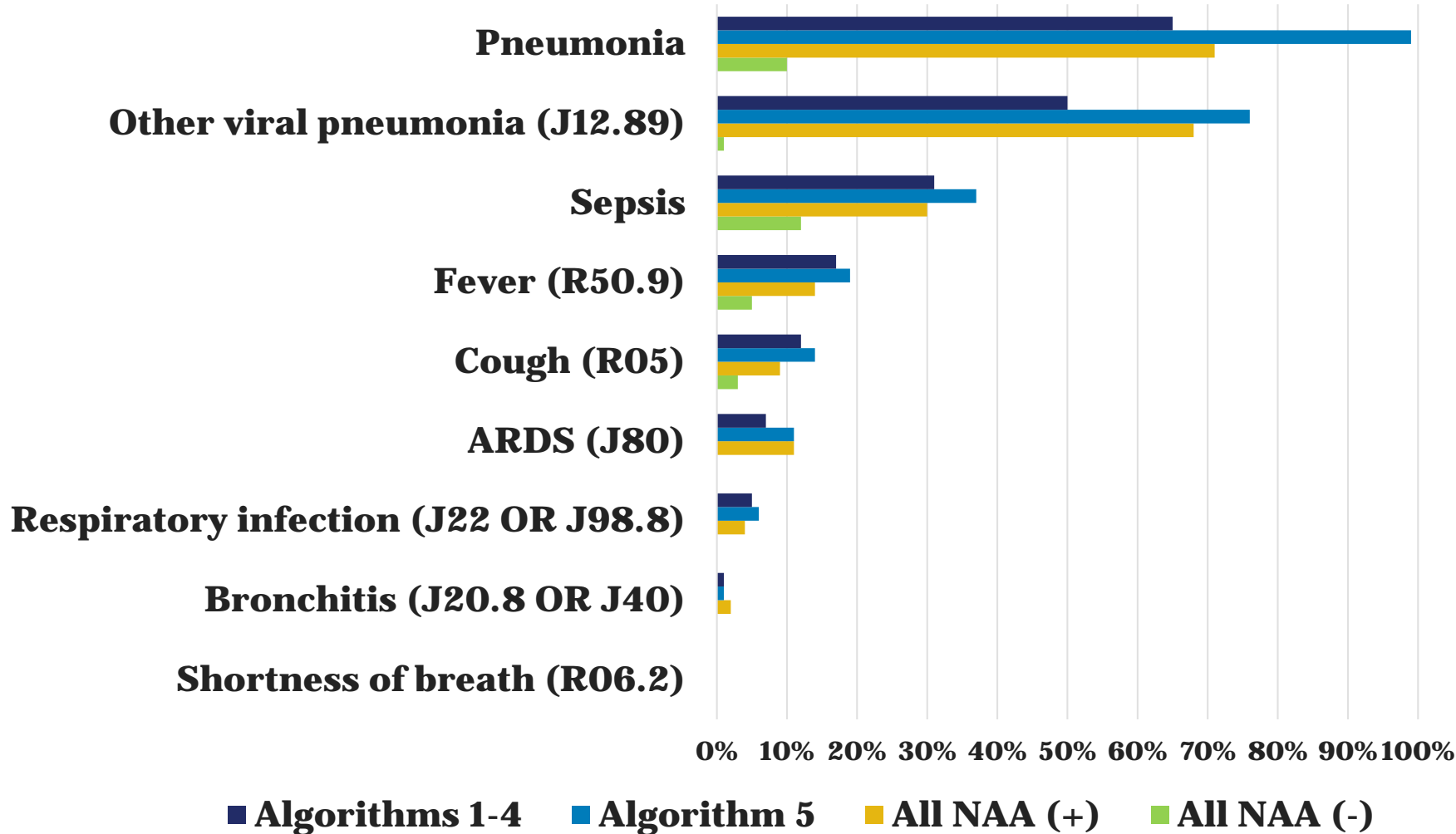
## Sex distribution



# Results: Cohort clinical characteristics

**Algorithm 1:** U07.1  
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## Symptoms

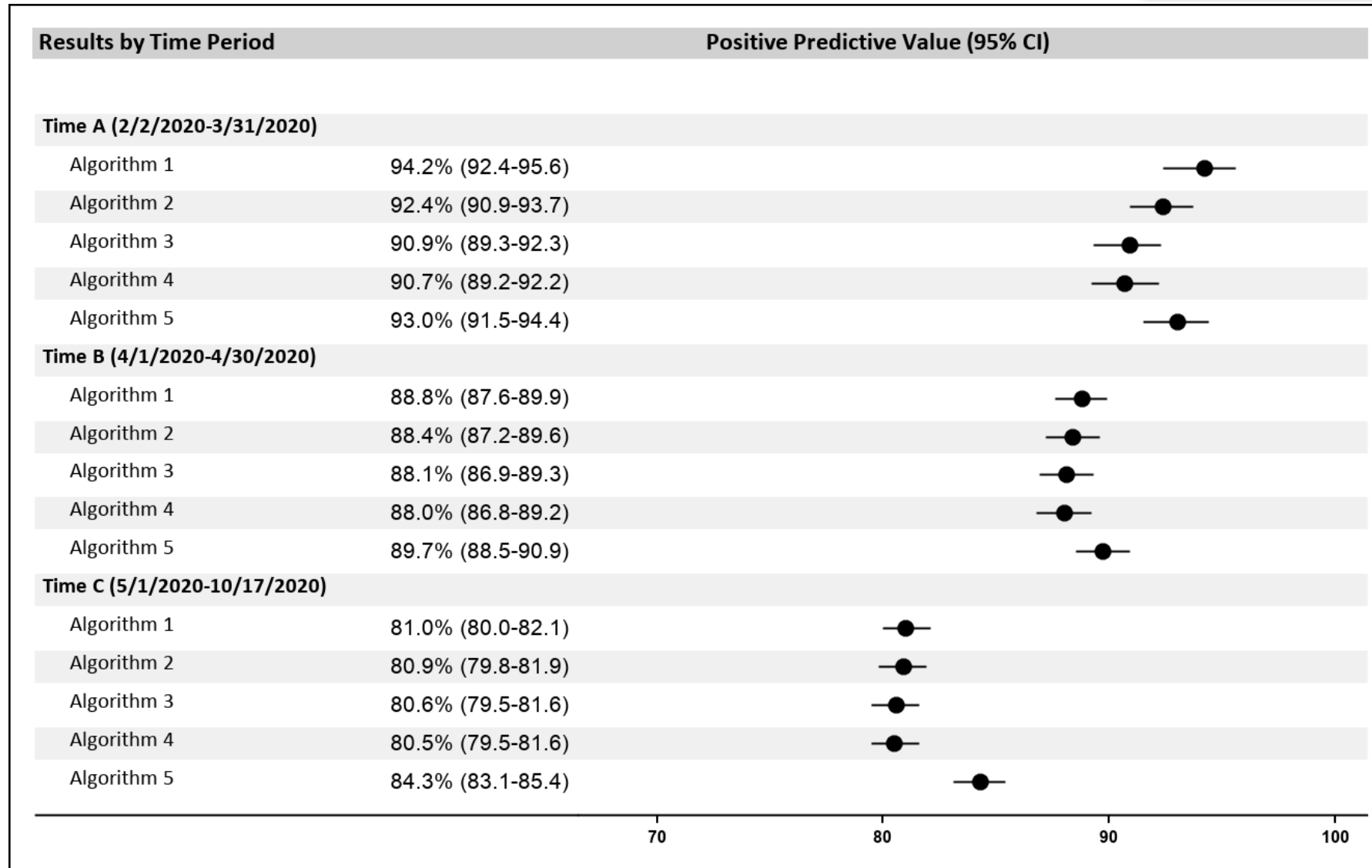


**Algorithm-based cohorts (1-4) and lab-based cohort** have similar symptoms

**Severe COVID-19 cohort (algorithm 5)** has a slightly higher prevalence of sepsis, fever, and cough

# Results: PPV

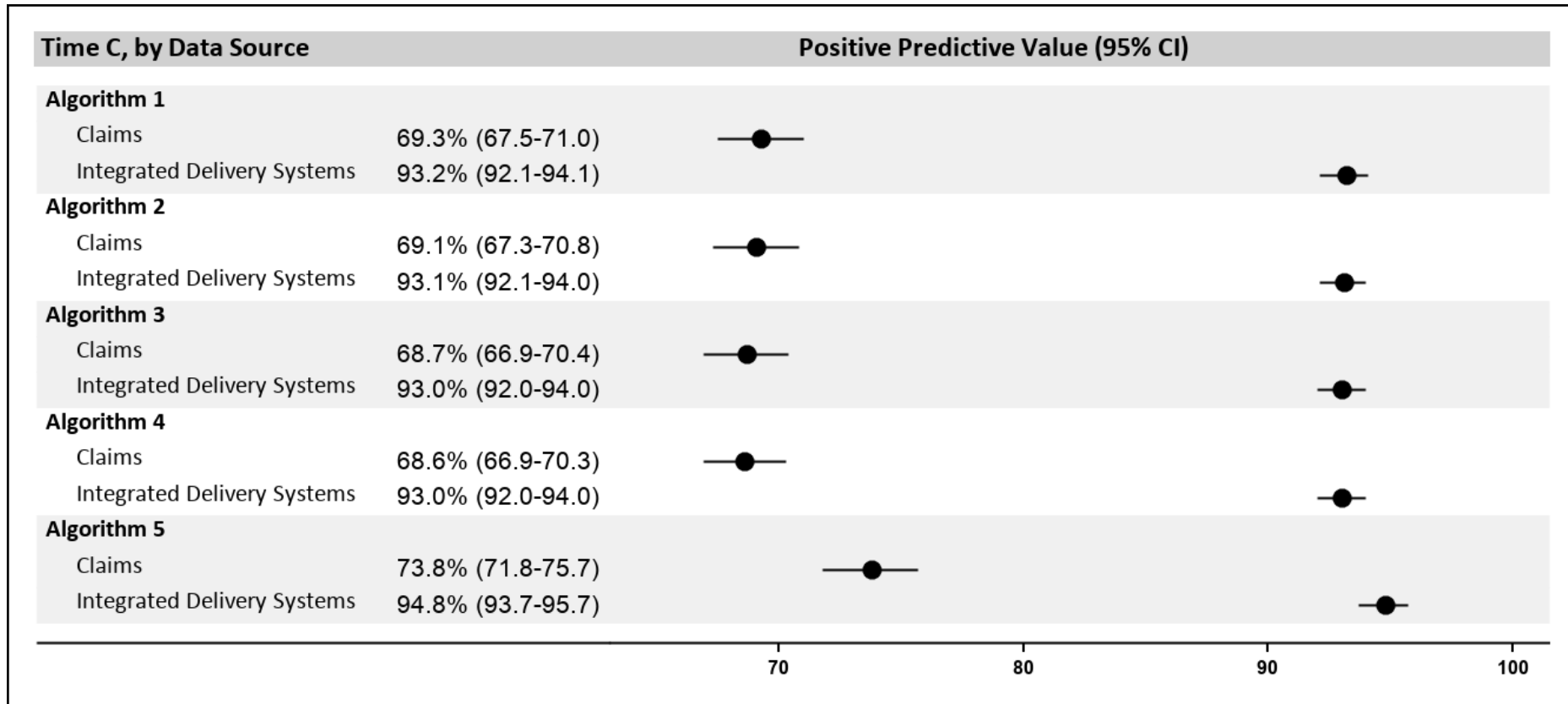
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**Aggregated across all data partners, PPV decreases from ~91% in Time A to ~81% in Time C**

# Results: PPV by data source

**Algorithm 1:** U07.1  
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**Overall PPV of 81% in Time C actually reflects:**

- **69% PPV among claims data partners**
- **93% PPV among integrated delivery systems**

# Results: Sensitivity

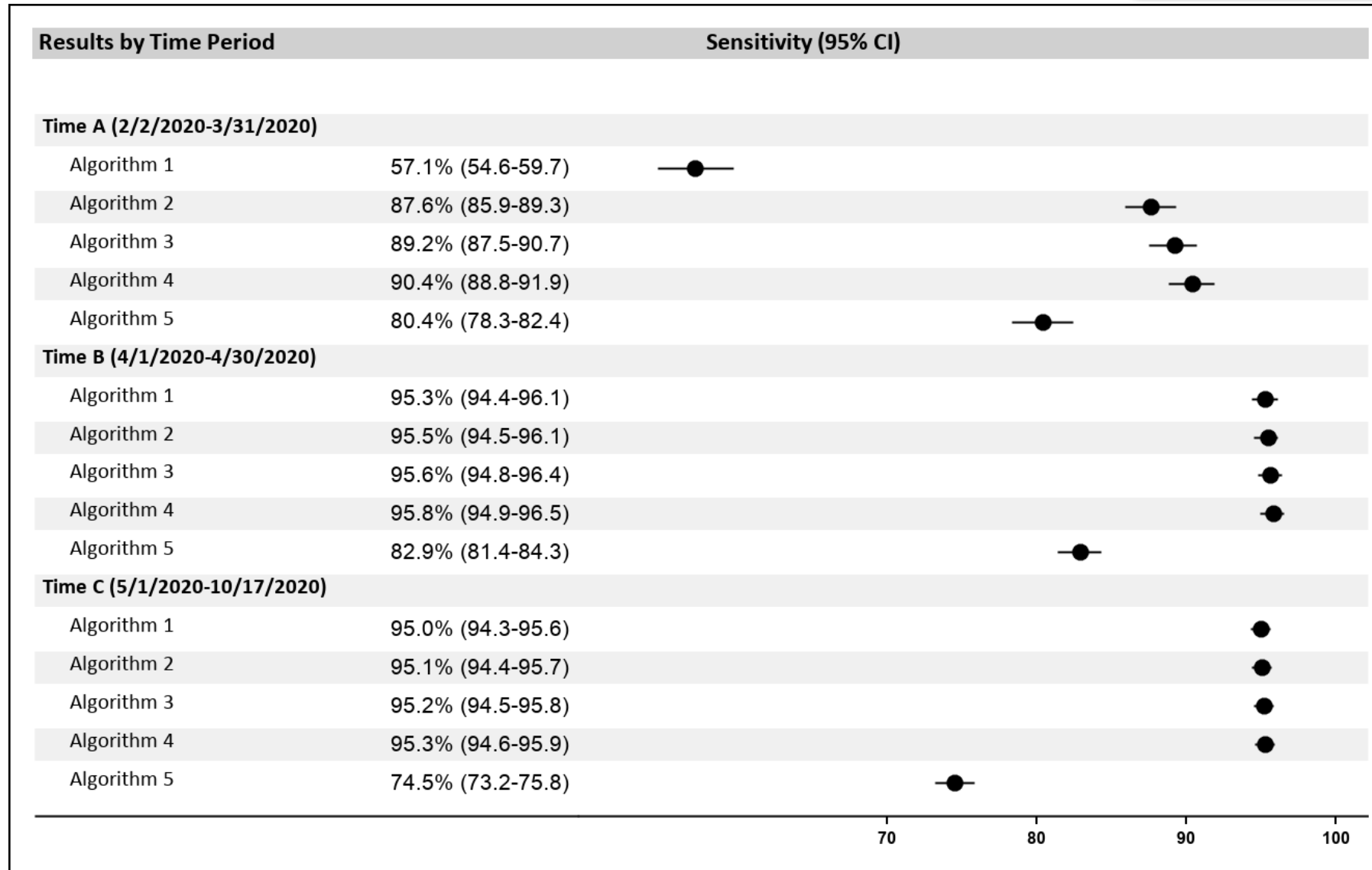
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## Time A:

- Lower sensitivity
- U07.1 did not yet exist

## Times B & C:

### Algorithms 1-4:

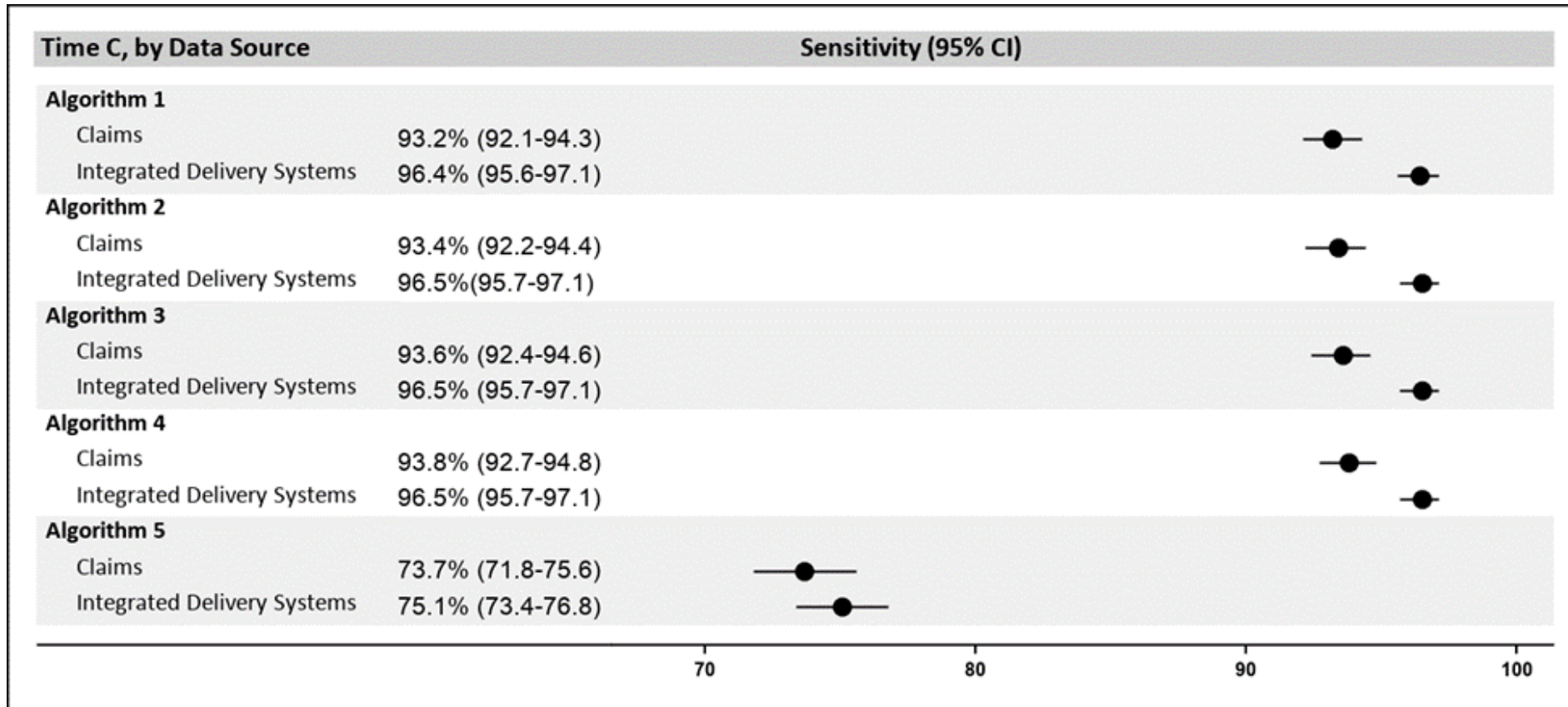
- Sensitivity consistent across algorithms
- Sensitivity stable at 95%

### Algorithm 5:

- Sensitivity substantially lower than for other algorithms.
- Declines over time

# Results: Sensitivity by data source

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**Algorithm 3:** U07.1 or B97.29 or B34.2  
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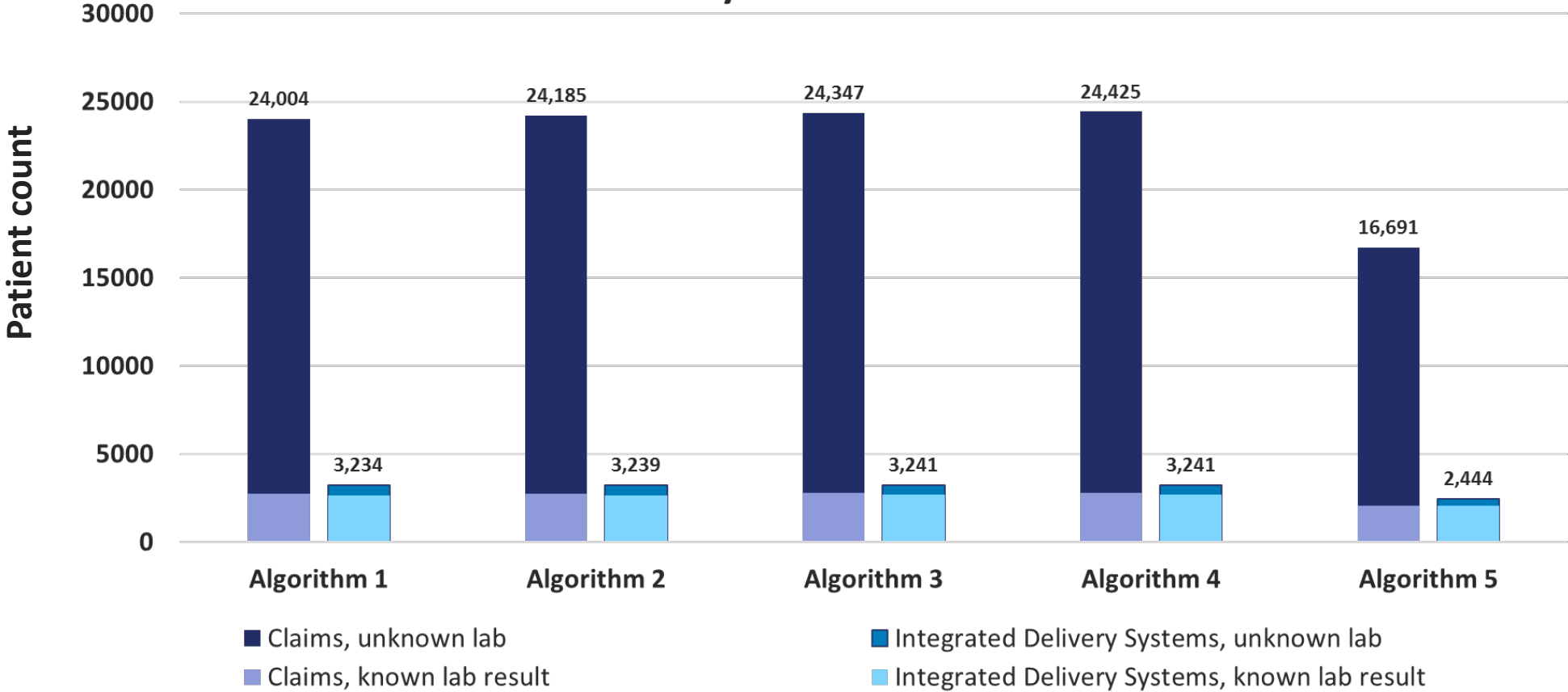


**Sensitivity is similar for claims and integrated delivery systems**



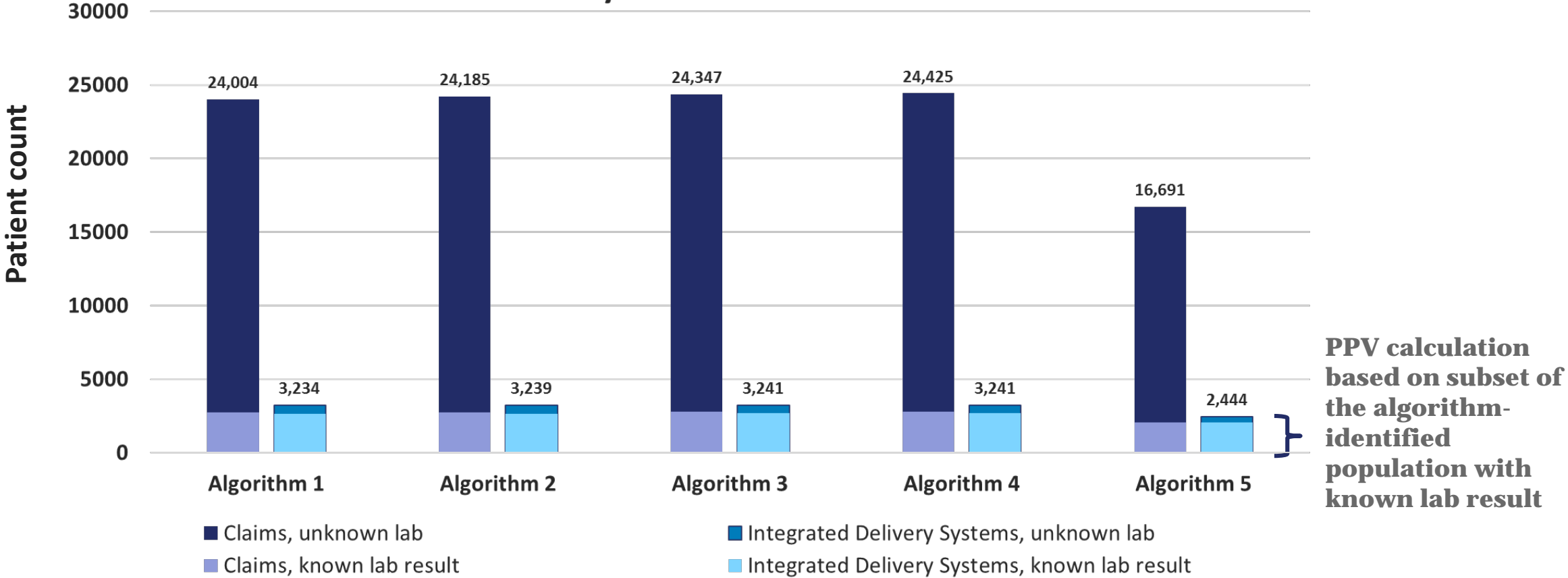
# Results: Interpretation

## Time C Algorithm Capture and Proportion with Known Lab Result, by Data Source



# Results: Interpretation

Time C Algorithm Capture and Proportion with Known Lab Result, by Data Source



# Results: Limitations

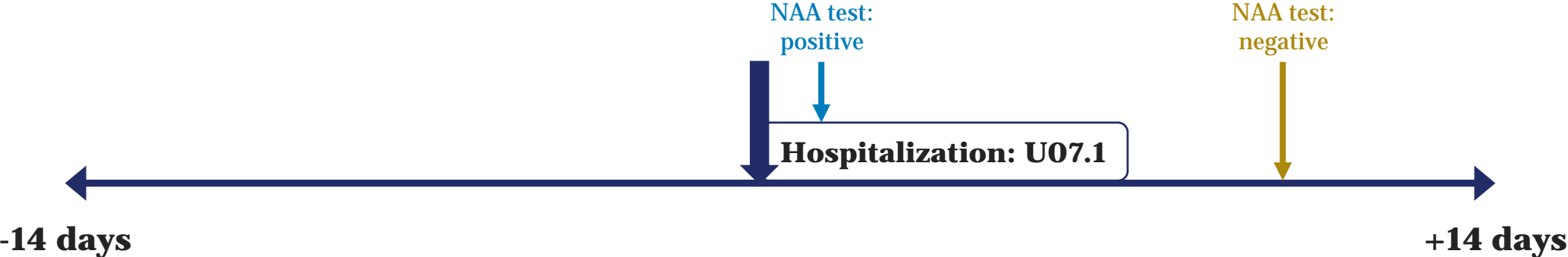
## Data capture among claims data partners:

- **Data missingness:** Claims data partners have **little inpatient lab data**; their lab results come from large national diagnostic laboratories that mostly process outpatient tests.
- **Not at random:** Among hospitalized patients with COVID-19 diagnoses, inpatient lab results are likely to be positive. **We believe we are missing some of these inpatient tests.**
- **Resulting bias:** In claims data, patients with a negative outpatient test before or after hospitalization for COVID-19 could be **misclassified as lab-negative** even with a positive inpatient test.

# Results: Limitations

**Claims: Capture of lab results differs by care setting**

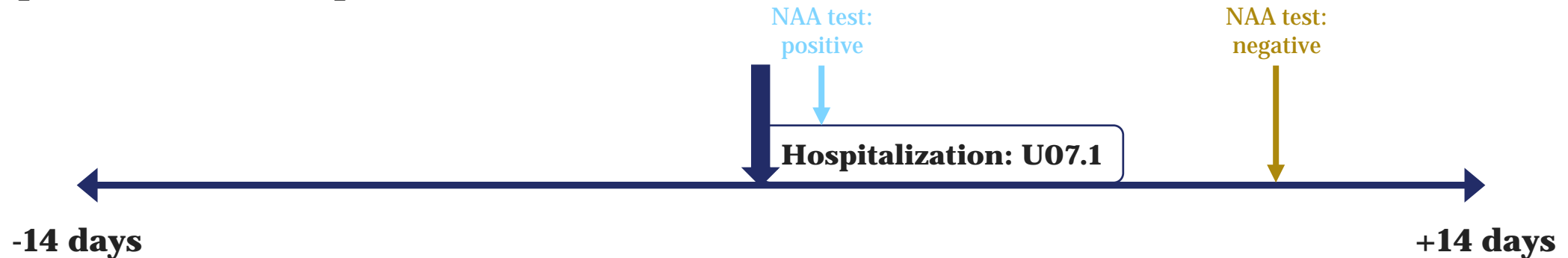
**Sample misclassified patient:**



# Results: Limitations

**Claims: Capture of lab results differs by care setting**

**Sample misclassified patient:**



**Inpatient tests are not observed;  
patient is classified as algorithm-positive and NAA-negative**

# Summary

## Findings:

- **U07.1** alone captured almost as many patients as the broader algorithms, with nearly identical PPV (Time C, all data partners: 81% PPV).
- **Observed PPV** of algorithms 1-4 declined across the three time periods.
  - Claims: Observed PPV declined from 93% to 69% over time.
  - Integrated delivery systems: PPV was >90% across all time periods.
- **Sensitivity** was stable at 95% for Times B and C.
- **Algorithm 5**, “severe COVID-19,” had a higher PPV and lower sensitivity.

# Summary

## Conclusions:

- PPV is based on more complete data for integrated delivery systems compared to claims.
- We believe that the observed differences in PPV by health plan type are partly attributable to differences in capture of inpatient laboratory data.
- With a PPV of 81% and sensitivity of 95%, we recommend using U07.1 for inpatient COVID-19 cohort identification in U.S. claims data when complete laboratory results are not available.

# Workgroup

## **Harvard Pilgrim Health Care Institute:**

Sheryl Kluberg

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## **U.S. Food and Drug Administration:**

Sarah Dutcher

Brian Kit

Monisha Billings

Michael Nguyen

Robert Ball



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HealthPartners

Kaiser Permanente Colorado

Kaiser Permanente Mid-Atlantic States

Kaiser Permanente Washington



Colorado  
Mid-Atlantic  
Washington



# Thank you

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