## NHSN Quarterly Validation Call For State HAI Coordinators

Friday, September 28, 2018

2:00pm - 3:00pm EST

## **Today's Agenda**

- Introduction
- Update 2018 HAI Validation Guidance and Toolkits
- Presentation Facility Selection for External Validation of HAI Data Reported to NHSN: Alternative Approach
- Presentation Data Validation in North Carolina 2018
- Question & Answer Session
- Wrap-up

## **NHSN HAI Validation Team**

- Suparna Bagchi, MSPH, DrPH, HAI Validation Lead
  - iyj9@cdc.gov
- Bonnie Norrick, MT(ASCP), EdM, CIC, CPHQ
  - ojd8@cdc.gov
- Jennifer Watkins, RN, BSN, MPH
  - nub7@cdc.gov

## **2018 Validation Guidance and Toolkits**

- 2018 External and Internal Validation Guidance and Toolkits are posted!
  - <u>https://www.cdc.gov/nhsn/validation/index.html</u>

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## **2018 External Validation Guidance and Toolkit**

- 2018 External Validation Guidance and Toolkit Updates:
  - Two methods of facility selection
  - Updated instructions, including NHSN screenshots
  - MRATs updated and reformatted
- 2018 Internal Validation Guidance and Toolkit Updates:
  - Addition of Data Quality checklists

## **MRAT Updates 2018 - Location**

#### NHSN Validation Guidance and Resources for 2018

- > For Reporting Facilities: 2018 Internal Validation Guidance and Toolkit
- For Auditors: 2018 External Validation Guidance and Toolkit
  - 2018 External Validation Guidance and Toolkit 🛃 [PDF 3 MB]

Medical Record Abstraction Tools (MRAT) and Instructions

- 2018 CLABSI Medical Record Abstraction Tool (MRAT) 🔂 [PDF 300 KB] (print-only)
  - 2018 Instructions for CLABSI MRAT 1 [PDF 300 KB] (print-only)
- 2018 CAUTI Medical Record Abstraction Tool (MRAT) 🛃 [PDF 300 KB]
  - 2018 Instructions for CAUTI MRAT 1 [PDF 300 KB] (print-only)

## MRAT Updates 2018 – New Field

**Case Determination** (A) Correctly Classified (B) Over-reported HAI (C) Underreported HAI If CLABSI was misclassified (over- or underreported) by facility, what was the reason? (I) General HAI definition misapplication (II) CLABSI criteria misapplied (IIa) Central Line not in > 2 days in an inpatient location on date of (la) Incorrect location of attribution (Ib) Date of event incorrect event (IIb) Missed CLABSI due to central line removed day of or day before (Ic) IWP set incorrectly the date of event (Id) RIT applied incorrectly (IIc) Missed CLABSI due to location transfer/discharge day of or day (Ie) Did not identify elements present in IWP before the date of event (If) POA/HAI applied incorrectly (IId) CLABSI incorrectly identified as secondary BSI (Ih) Other (IIe) Secondary BSI incorrectly identified as a primary CLABSI (III) Additional Reasons (IIf) Other (IIIa) Missed case finding/failure to review positive specimen/culture (IIIb) Clinical over-rule (IIIc) Used outdated criteria (IIId) No positive blood specimen in chart (Ille ) Other - -

July 2018

## **Data Quality Checklists - 2018**

#### Appendix G: Data Quality Checklist - CLABSI/CAUTI Data

This checklist is intended to ensure completeness and accuracy of CLABSI and CAUTI data entered into NHSN and can be used at acute care hospitals, long term acute care facilities, critical access hospitals, and inpatient rehabilitation facilities.

Summary Denominator Data					
Indicator	Description/Action		Validated		
i) Missing summary data	Verify that summary data has be	en entered for the			
	location and month/year. (Go to	NHSN Application ->			
	Alerts -> Missing Summary Data)				
ii) Missing denominator variables (Incomplete	Verify that all mandatory/require	ed fields are			
summary data)	completed, and that "Report No	Events" is checked, if			
	appropriate. (Go to NHSN Applica	ation -> Alerts ->			
	Incomplete Summary Data)				
<li>iii) Verify denominator data accuracy:</li>	Generate Rate Tables to display	location for multiple	months.		
	location and month in a table f	Event Data Entry			
		Indicator	-	Description/Action	Validated
		i) All CLABSI and CAI	JTI events reported	Verify that all CLABSI and CAUTI events have been	
				reported.	
				Go to NHSN Application -> Analysis -> Reports ->	
				Device-Associated (DA) Module -> Central Line-	
				Associated BSI -> Line Listing - All CLAB Events	
				OR	
				NHSN Application -> Analysis - Reports -> Device-	
				Associated (DA) Module -> Urinary Catheter-	
				Associated UTI -> Line Listing - All CAU Events	
		ii) Missing numerato	r variables (Incomplete events	) Verify that all mandatory/required data fields	
				(marked with an *, **, or > on the event form) are	
				completed. (Go to NHSN Application -> Alerts ->	
				Incomplete Events, Event Type: BSI/UTI)	
		iii) Confirm that date	e of event occurred on or after	If the event did not occur on or after the third	

## **Today's Speakers**

- Suparna Bagchi, MSPH, DrPH
  - HAI Validation Lead
  - CDC NHSN Protocol and Validation Team
  - <u>iyj9@cdc.gov</u>
- Savannah Carrico, MPH
  - HAI Epidemiologist, SHARPPS Program
  - North Carolina Division of Public Health
  - <u>savannah.carrico@dhhs.nc.gov</u>

## Facility Selection for External Validation of HAI Data Reported to NHSN: Alternative Approach

Suparna Bagchi, MSPH, DrPH HAI Validation Lead Protocol and Validation Team September 28, 2018

## **Objectives**

- Review the methods of facility selection in NHSN External Validation Guidance
- New method of facility selection in 2018 Guidance
- Comparison of facility selection methods
- Recommended data analysis and summarization

# **Facility Selection Method 1**

- Targeted sampling: facility specific predicted events and SIR
- Facilities are sorted based on predicted number of events
- Top third of facilities (tertiles):
  - Targeting and prioritization
  - Facility specific SIR relative to median SIR for the top tertile of the facilities
- SIR does not estimate absolute burden of HAIs in a facility
- Ratio of observed/predicted events
- Focuses on larger (higher burden facilities), excludes smaller facilities where underreporting could be a potential problem

## **Method 2: Alternative Approach**

- Underreporting of HAI remains primary concern
- Cumulative Attributable Difference (CAD) approach
- CAD = Observed HAIs (Predicted HAIs \* SIR Goal)
   CAD = Observed events Predicted events
- Facilities could have both positive and negative CAD values
- Facilities reporting zero or very few events: negative CAD value
- Prioritization based on highest negative CAD values can help assess the data accuracy among facilities with high predicted and very few or no reported events during a time frame

## **Comparison of Facility Selection Methods**

	Method 1 - Prioritizing Facilities with Highest Likelihood of Event Occurrence	Method 2 - Cumulative Attributable Difference (CAD) Approach
Facility Selection criteria	<ul> <li>Based on highest likelihood of event occurrence.</li> </ul>	<ul> <li>Based on difference of predicted and observed number of events.</li> </ul>
Which type of facilities are selected?	<ul> <li>Larger facilities with higher predicted/expected number of events are more likely to be selected</li> </ul>	<ul> <li>Prioritization focuses on facilities with negative values of difference, primarily under- reporters</li> </ul>

## **Comparison of Facility Selection Methods**

	Method 1 - Prioritizing Facilities with Highest Likelihood of Event Occurrence	Method 2 - Cumulative Attributable Difference (CAD) Approach
Ranking algorithm	<ul> <li>SIR metric is a ratio of and is subject to variability</li> <li>A small facility with low predicted volume of events with even one observed event could lead to a high SIR value.</li> </ul>	<ul> <li>Cumulative attributable difference (CAD)</li> <li>CAD metric is robust, stable and reflects the true facility HAI burden</li> </ul>
Which method should my state use?	<ul> <li>No prior validation, use Method 1 to determine errors in HAI misclassification</li> <li>If already aware of underreporting concerns - select Method 2</li> </ul>	<ul> <li>Previous validation history that have identified underreporting as a potential concern would benefit additionally with this method</li> </ul>

## **CAD Method of Facility Selection**

- Generate new datasets in NHSN
- After successful dataset generation, navigate to Analysis
- Navigate to the SIR report of interest
- Export Analysis Data Set screen export to an Excel spreadsheet
- Exported SIR report file will display multiple levels of aggregation
- In Excel, select the aggregation level that provides a facilityspecific SIR for all validation locations

## **Facility SIR Level View**

4	1.71106	1030	0.1252	2.338	0.743, 5.639	IN:ACUTE:CC:M_PE	D	
2	3.18614	2824	0.5558	0.628	0.105, 2.074	IN:ACUTE:CC:NS		
7	9.78433	7188	0.3846	0.715	0.313, 1.415	IN:ACUTE:CC:NURS	SIRs for each	location types
9	17.3151	15646	0.0327	0.52	0.253, 0.954	IN:ACUTE:CC:S		
2	3.26009	2134	0.531	0.613	0.103, 2.027	IN:ACUTE:CC:T		
	0.75000	070						
3	5.18805	4183	0.3493	0.578	0.147, 1.574		HOSP-GEN	
1	9.06437	7682	0.0013	0.11	0.006, 0.544		HOSP-GEN	
2	7.57817	6272	0.0235	0.264	0.044, 0.872		HOSP-GEN	
0	0.42346	562					HOSP-GEN	
0	0.7199	823					HOSP-GEN	"T IIS IS THE LEVEL TO EVALUATE"
2	1.0873	1253	0.3934	1.839	0.308, 6.077		HOSP-GEN	Fa ility-specific SIRs combining all location types
0	0.44531	591					HOSP-GEN	
1	1.05264	933	1	0.95	0.048, 4.685		HOSP-GEN	
13	18.5196	15267	0.1921	0.702	0.390, 1.170		HOSP-GEN	
0	0.15574	232					HOSP-GEN	
1	1.52725	1760	0.7659	0.655	0.033, 3.229		HOSP-GEN	
3	1.57237	1812	0.2846	1.908	0.485, 5.193		HOSP-GEN	
2	3 60045	21/0	0.4283	0 555	0.003 1.8 00		HOSP-CEN	
1	1.5876	1034	0.7334	0.63	0.032, 3.1 CC_N		HOSP-GEN	facility and location types
1	8.26831	7012	0.0026	0.121	0.006, 0.5 CC		HOSP-GEN	
0	0.79606	670			CC_N		HOSP-GEN	

## **Calculate the 75<sup>th</sup> Percentile Value of numPred**

infCount	numPred	numciday	SIR_pval	SIR	sir95ci	locationT	locCDC	orgID	facType	
13	18.51959	15267	0.1921	0.702	0.390, 1.17	0		100008	HOSP-GEN	
22	15.32671	9910	0.1034	1.435	0.922, 2.13	8		100030	HOSP-CHLD	
10	9.736101	8387	0.8926	1.027	0.522, 1.83	1				
8	9.542312	7958	0.6509	0.838	0.389, 1.5					
1	9.064373	7682	0.0013	0.11	0.006, 0.	5, 0. Sort the facilities in the descending				
2	7.578169	6272	0.0235	0.264	0.044, 0.	ord	er of ni	umber (	of predicted	d infections
7	5.689505	4581	0.5585	1.23	0.538, 2.		/ -			
4	5.504663	<b>1072</b>	0.000	01727	0.231, 1.		(numPi	red) an	d compute	the 75 <sup>m</sup>
2	3.159258	2784	0.5651	0.633	0.106, 2.		percer	ntile va	lue of the v	ariable
2	2.437844	2304	0.8601	0.82	0.138, 2.		Percei			
0	1.945079	1724	0.143	0	, 1.540			nu	ImPred	
3	1.572374	1812	0.2846	1.908	0.485, 5.1.					
1	1.527251	1760	0.7659	0.655	0.033, 3.22	9		100010	HOSP-GEN	
2	1.329405	1357	0.5333	1.504	0.252, 4.97	0		100032	HOSP-GEN	
0	1.242188	1101	0.2888	0	, 2.412			100049	HOSP-GEN	
2	1.087298	1253	0.3934	1.839	0.308, 6.07	7		100005	HOSP-GEN	
1	1.052644	933	1	0.95	0.048, 4.68	5		100007	HOSP-GEN	
1	0.915007	934						100040	HOSP-GEN	
0	0.745198	989						100026	HOSP-GEN	
0	0.719899	823						100004	HOSP-GEN	
2	0.669096	888						100023	HOSP-GEN	

## **Selection of Facility Sampling Frame**

infCount	numPred	numclday	SIR_pval	SIR	sir95ci	locationTy	locCDC	orgID	facType
13	18.51959	15267	0.1921	0.702	0.390, 1.17	70		100008	HOSP-GEN
22	15.32671	9910	0.1034	1.435	0.922, 2.13	38		100030	HOSP-CHL
10	9.736101	8387	0.8926	1.027	0.522, 1.83	31		100014	HOSP-GEN
8	9.542312	7958	0.6509	0.838	0.389, 1.59	92		100046	HOSP-GEN
1	9.064373	/682	0.0013	0.11	0.006, 0.54	14		100001	HOSP-GEN
2	7.578169	6272	0.0235	0.264	0.044, 0.87	72		100002	HOSP-GEN
7	5.689505	4581	0.5585	1.23	0.538, 2.43	34		100022	HOSP-GEN
4	5.504663	4879	0.558	0.727	0.231, 1.75	53		100027	HOSP-GEN
2	3.159258	2784	0.5651			-			OSP-GEN
2	2.437844	2304	0.8601	75 <sup>th</sup> p	ercentile v	alue of n	ImPred= 5	5.5. Select	P-GEN
0	1.945079	1724	0.143	faciliti	ies with n	umPred >5	5 Only f	acilities in	P-GEN
3	1.572374	1812	0.2846	rod k			no includ	ad in the	P-GEN
1	1.527251	1760	0.7659	read	ox (nume	rea >5.5) (	are includ	ea in the	P-GEN
2	1.329405	1357	0.5333	sar	sampling frame for targeted validation.				
0	1.242188	1101	0.2888						3P-GEN
2	1.087298	1253	0.3934	1.839	0.308, 6.07	77		100005	HOSP-GEN
		000		0.05	0.040 4.00	DC		400007	11000 051

## **Compute the CAD Values for Sampling Frame**

- Variable infCount
  - Pooled total observed events from all validation locations, for the timeframe of validation for each facility selected in sampling frame
- Insert a column (CAD) next to the numPred
- Compute CAD as difference: infCount numPred
- Could generate all negative, positive and negative, all positive

## Sort the Facilities by CAD Values

	-	-	-	-			-			•	
infCount	numPred	CAD	numclday	SIR_p	val	SIR	sir95ci	locationTy	locCDC	orgID	facType
1	9.064373	-8.06437	7682	0						100001	HOSP-GEN
2	7.578169	-5.57817	6272		Cor	npute the (	CAD values	for facilitie	s in the	100002	HOSP-GEN
13	18.51959	-5.51959	15267		san	npling fram	e.			100008	HOSP-GEN
8	9.542312	-1.54231	7958		Sor	t the CAD v	alues in de	scending o	rder	100046	HOSP-GEN
4	5.504663	-1.50466	4879		(hig	hest negat	ive on the	top). If the		100027	HOSP-GEN
10	9.736101	0.263899	8387		san	npling fram	e has great	er than 15		100014	HOSP-GEN
7	5.689505	1.310495	4581		faci	lities, selec	t the top 1	5 facilities.		100022	HOSP-GEN
22	15.32671	6.673286	9910	0.	-					100030	HOSP-CHLD

### Facility Selection: If Sampling Frame > 30 Facilities

- Divide the total facilities in the sampling frame into two strata:
  - Stratum 1: Includes all facilities in the sampling frame that had zero reported pooled observed events for the validation time frame
  - Stratum 1: will generate all negative CAD values
  - Stratum 2: includes all facilities in the sampling frame with non-zero reported pooled observed events for the validation time frame
  - Stratum 2: could generate positive and negative CAD values

## **Stratum 1: Facilities with Zero Reported Events**

- All CAD values will be negative.
- Highest negative values: facilities with greater predicted and zero events reported
- Sort them in descending order of negative values of CAD
- Facilities with the highest negative CAD value should be at the top
- Select the first 15 facilities from Stratum A.

## **Stratum 2: Facilities with Non-zero Reported Events**

- CAD values could be positive or negative
- Highest negative values: facilities with greater predicted and zero events reported
- Sort them in descending order of negative values of CAD
- Facilities with the highest negative CAD value should be at the top
- Select the first 15 facilities from Stratum B



- If value > 1, then use the value corresponding to 75<sup>th</sup> percentile, otherwise value = 1
- Create a subset of facilities in state with predicted events greater than the threshold



## **Medical Record Selection: CAD Approach**

- Before requesting medical records: download ("freeze") data
- Request facilities to send line lists of candidate HAI events
- For facilities with reported events in validation locations:
  - Events reported to NHSN in the validation time frame (select all)
  - Randomly select additional medical records for a total of 40 medical records for candidate cases.
- For facilities with no reported event in validation locations:
  - Randomly select 40 medical records for review for each HAI candidate event.

## **Recommended Data Summary**

	Auditor Det		
Facility	Events	Not Events	
Events reported	True Positive (a)	False Positive (b) Over reports	(a+b)
Events not reported	False Negative (c) Missed events	True Negative (d)	(c+d)
	(a+c)	(b+d)	Total

- True positive (a): facility identified and reported the events and auditor agreed
- True negative (d): facility did not identify/report event and auditor agreed
- False negative (c): facility did not identify/report event and auditor disagreed (MISSED)
- False positive (b): facility identified and reported the events and auditor disagreed (OVER REPORT)

## **Recommended Data Analysis**

	Auditor De		
Facility	Events	Not Events	
Events reported	True Positive (a)	False Positive (b)	(a+b)
Events not reported	False Negative (c)	True Negative (d)	(c+d)
	(a+c)	(b+d)	Total

- Sensitivity: Ability of a test to correctly identify those with the disease (true positive rate) = a/(a+c)
- Specificity: Ability of the test to correctly identify those without the disease (true negative rate) = d/(b+d)
- Positive Predictive Value: Proportion of individuals who test positively (a+b) AND truly have the disease (a) = a/(a+b)
- Negative Predictive Value: Proportion of individuals who test negatively (c+d) AND truly do not have the disease (d) = d/(c+d)

## **Reasons for Misclassification**

- For each misclassified case, list the reasons for errors in reports
- Compute proportion of each error type identify gaps, training opportunities

#### **Reasons for under-reported CDI events**

- Incorrect understanding of protocol definition (n1)
- Laboratory records missed (n2)
- Reason ....

**Total Under-reported events** 

#### **Reasons for over-reported CDI events**

- Incorrect specimen (n1)
- Duplicate record (n2)
- Reason ....

#### **Total Over-reported events**

# **Summary and Recommendations**

- Both facility selection methods use a targeted approach
- Generalizability is still limited
- Select the method as deemed appropriate
- Compare same HAI validated previously validated using alternative method
- Feedback on implementation: challenges and successes

### Questions !

### Suparna Bagchi – <u>iyj9@cdc.gov</u> Jennifer Watkins – <u>nub7@cdc.gov</u> Bonnie Norrick – <u>ojd8@cdc.gov</u>

For more information, contact CDC 1-800-CDC-INFO (232-4636) TTY: 1-888-232-6348 www.cdc.gov

The findings and conclusions in this report are those of the authors and do not necessarily represent the official position of the Centers for Disease Control and Prevention.





## **Data Validation in North Carolina 2018**

Savannah Carrico, MPH HAI Epidemiologist September 28, 2018

## **Outline**

- I. Importance of Data Validation
- II. Hospital Selection Method: SIR and CAD
- **III.** Results of North Carolina's CDI and CLABSI validations

## **Importance of Data Validation**

• Non punitive validation

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- Engages health care facilities in accurate data collection methods

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# **Importance of Data Validation**

- Non punitive validation
- Engages health care facilities in accurate data collection methods
- The goal identify the true burden of HAIs
- Accurate data in NHSN allows for comparable data
- Opportunity for facilities and validators to discuss HAI prevention and response

## **SHARPPS Program Data Validation**

- The North Carolina Surveillance for Healthcare-Associated Resistant Pathogens Patient Safety (SHARPPS) Program has been performing data validation HAIs since 2015
- SHARPPS performs data validation without funding
- Since 2015 CLABSI, CDI, CAUTI, and MRSA have been validated

# **Selecting a Sample**

- There 93 Acute Care Hospitals in North Carolina
- CDC recommends 18 facilities be selected for states that have 21-149 hospitals
- Want to select hospitals that represent the state
- Selecting those that would benefit the most from data validation
- Must select hospitals without introducing bias



- Want to avoid asking facilities to self-select
- Want to select representative facilities
- Want to target facilities that would benefit the most

# **CDC** methodology









# **Results**

The majority of facilities were:

- All 18 facilities were in the top tertile
  - Highest number of predicted events
- In urban areas
  - North Carolina is 80% rural
  - 67 of 93 hospitals are in rural counties
- Trauma centers
  - Affiliated with major medical schools
  - Experience high volume of higher acuity patients

https://www.nccommerce.com/lead/research-publications/the-lead-feed/artmid/11056/articleid/123/rural-center-expands-its-classification-of-north-carolina-counties

Positives	Considerations
Focuses on high-burden facilities	

#### **Positives**

- Focuses on high-burden facilities
- Acknowledges potential for over- and underreporting within the top third of facilities by stratifying by Median SIR

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• Excludes facilities with < 1 Predicted Event

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- Excludes smaller facilities

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- Weighted selection of facilities (Top Tertile only)

#### **Positives**

- Focuses on high-burden facilities
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- Excludes facilities with < 1 Predicted Event
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- SIR doesn't estimate the absolute burden of HAIs on a facility because it is a ratio of observed to predicted infections

#### **Positives**

- Focuses on high-burden facilities
- Acknowledges potential for over- and underreporting within the top third of facilities by stratifying by Median SIR

- Excludes facilities with < 1 Predicted Event
- Excludes smaller facilities
- Weighted selection of facilities (Top Tertile only)
- SIR doesn't estimate the absolute burden of HAIs on a facility because it is a ratio of observed to predicted infections
- Relies on accurate risk-adjustment of facilities

CAD = Observed # HAIs - (Predicted # HAIs \* SIR Goal)

- Calculated even if the number of predicted events is < 0 (Unlike SIR)</li>
- Represents the number of infections needed to be prevented to reach SIR goal
- The CAD can be used to identify facilities that would benefit the most from data validation
- NOT used for interfacility comparison

# **CAD** methodology





Positives	Considerations
<ul> <li>Captures facilities with &lt; 1 predicted event</li> </ul>	

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<ul> <li>Includes equal sample of facilities with 0 HAI events and &gt; 0 HAI events</li> </ul>	

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<ul> <li>Includes equal sample of facilities with 0 HAI events and &gt; 0 HAI events</li> </ul>	
<ul> <li>CAD accurately reflects absolute HAI Burden on a hospital</li> </ul>	

Positives	Considerations
• Captures facilities with < 1 predicted event	
<ul> <li>Includes equal sample of facilities with 0 HAI events and &gt; 0 HAI events</li> </ul>	
<ul> <li>CAD accurately reflects absolute HAI Burden on a hospital</li> </ul>	
<ul> <li>Potential to identify facilities with excellent prevention</li> </ul>	

#### **Positives**

- Captures facilities with < 1 predicted event
- Includes equal sample of facilities with 0 HAI events and > 0 HAI events
- CAD accurately reflects absolute HAI Burden on a hospital
- Potential to identify facilities with excellent prevention

#### **Considerations**

• Relies on accurate risk-adjustment of facilities

# **Choosing a methodology**

- The current method (SIR) has its pros but there are several considerations
- The CAD method:
  - addresses the considerations of the SIR method
  - selected representative facilities of North Carolina
  - captured both under and overreporting facilities
  - method was chosen as the selection method

# **CDI Results**

- 20 Facilities Validated
  - 13 ACHs
  - 2 LTACHs
  - 5 IRFs
- 1542 records validated
- 1 validator per record
- 95 % Facility and Validator Agreement
- 5% (79 records) not reported in NHSN that should have been
  - 87% (69 records) of these records were community onset

# **CLABSI Results**

- 12 Facilities Validated
- 293 Records Reviewed
- 2 validators per record
- 98% Agreement between facility and validators
- 94% Agreement between validators
- 2% (6 records) were discrepant
  - 1 record was misclassified as not a CLABSI by the facility
  - 6 records were misclassified as CLABSIs by the facility
    - 3 records Secondary to other infections
    - 2 records were not in reporting locations
    - 1 record had no central line



# **In Summary**

- Primary goal is to capture generalizable and representative data for the state
- The high agreement between facilities and validators suggests a thorough understanding of the NHSN surveillance definitions for CDIs and CLABSIs
- Future validations would be beneficial for all HAIs

### Acknowledgements

The North Carolina Surveillance for Healthcare-Associated Resistant Pathogens Patient Safety Program would like to acknowledge and appreciate all participating healthcare facilities in North Carolina



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# **Questions??**

# Wrap-Up

- Next Quarterly Call scheduled for Friday, January 11, 2019 from 2-3pm EST
- Is there anyone else we should invite? Please forward their name and email to Bonnie Norrick ojd8@cdc.gov.
- If you are interested in sharing your validation experience on a Quarterly Validation Call, please reach out to the NHSN HAI Validation Team
## Thank You!

## Please Join us for the Next

## NHSH Quarterly Validation Call for HAI Coordinators Friday, January 11, 2019 2:00pm—3:00pm EST For Questions Email <u>NHSN@cdc.gov</u>

For more information, contact CDC 1-800-CDC-INFO (232-4636) TTY: 1-888-232-6348 www.cdc.gov

The findings and conclusions in this report are those of the authors and do not necessarily represent the official position of the Centers for Disease Control and Prevention.

