

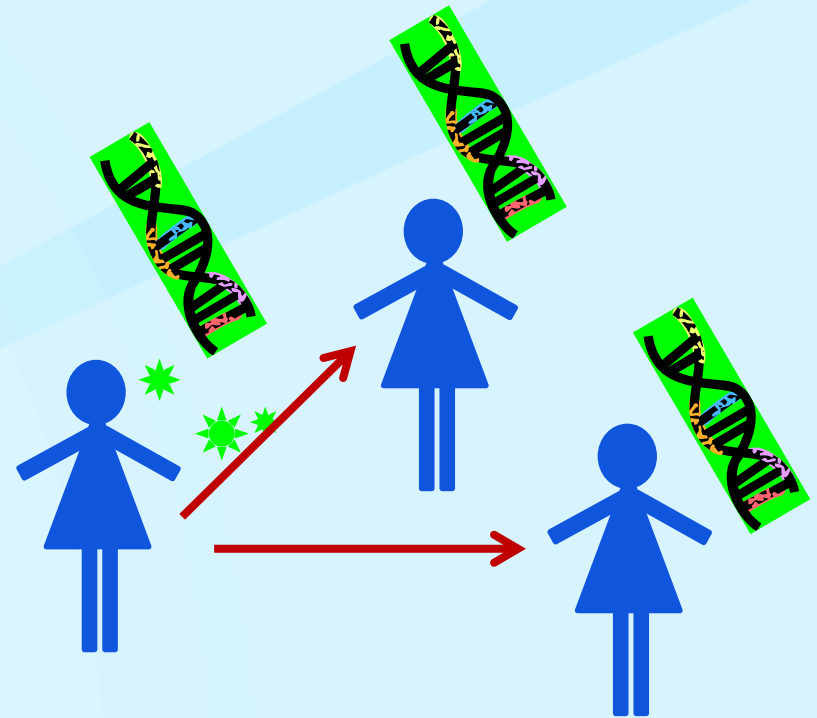
Using Routinely Collected Surveillance Data to Predict TB Outbreaks in the United States

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TB genotype background

- ❑ Examines genetic variation to determine relatedness of TB strains from different cases
- ❑ Only performed with culture-positive cases
- ❑ Outbreaks occur when highly infectious TB patients transmit TB
 - “Same” place
 - “Same” time
 - Secondary cases have same genotype

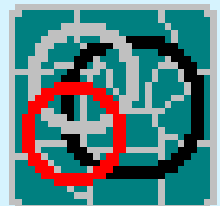


Methods overview

- ❑ **Genotyping with spoligotyping + 12-locus MIRU-VNTR**
- ❑ **SaTScan**
 - To identify TB clusters (to define our analytic cohort)
- ❑ **Decision Tree Analysis**
 - To identify clusters at high risk of becoming outbreaks

SaTScan

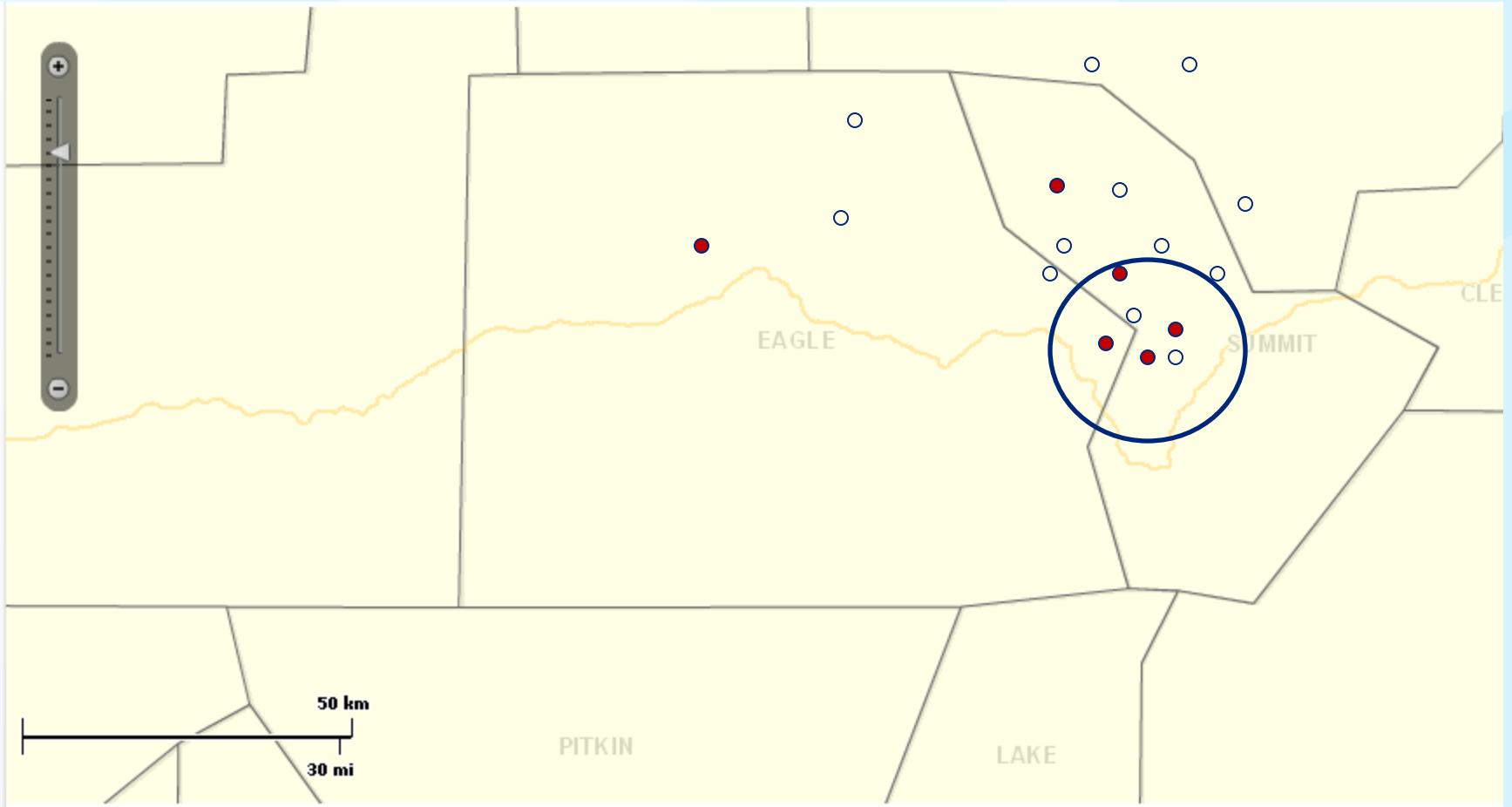
- ❑ **GIS software standard**
 - Performing geographical surveillance of disease
 - Detecting spatial disease clusters
- ❑ **Poisson-based model**
- ❑ **Monte Carlo simulation used to compute p-value for each cluster**



SaTScan parameters

- ❑ TB cases with genotyping results reported in U.S. from 2006 through 2010**
- ❑ Search within 50 km radius**
- ❑ Significant and non-significant**

Sample SaTScan map



How we defined an incident TB cluster

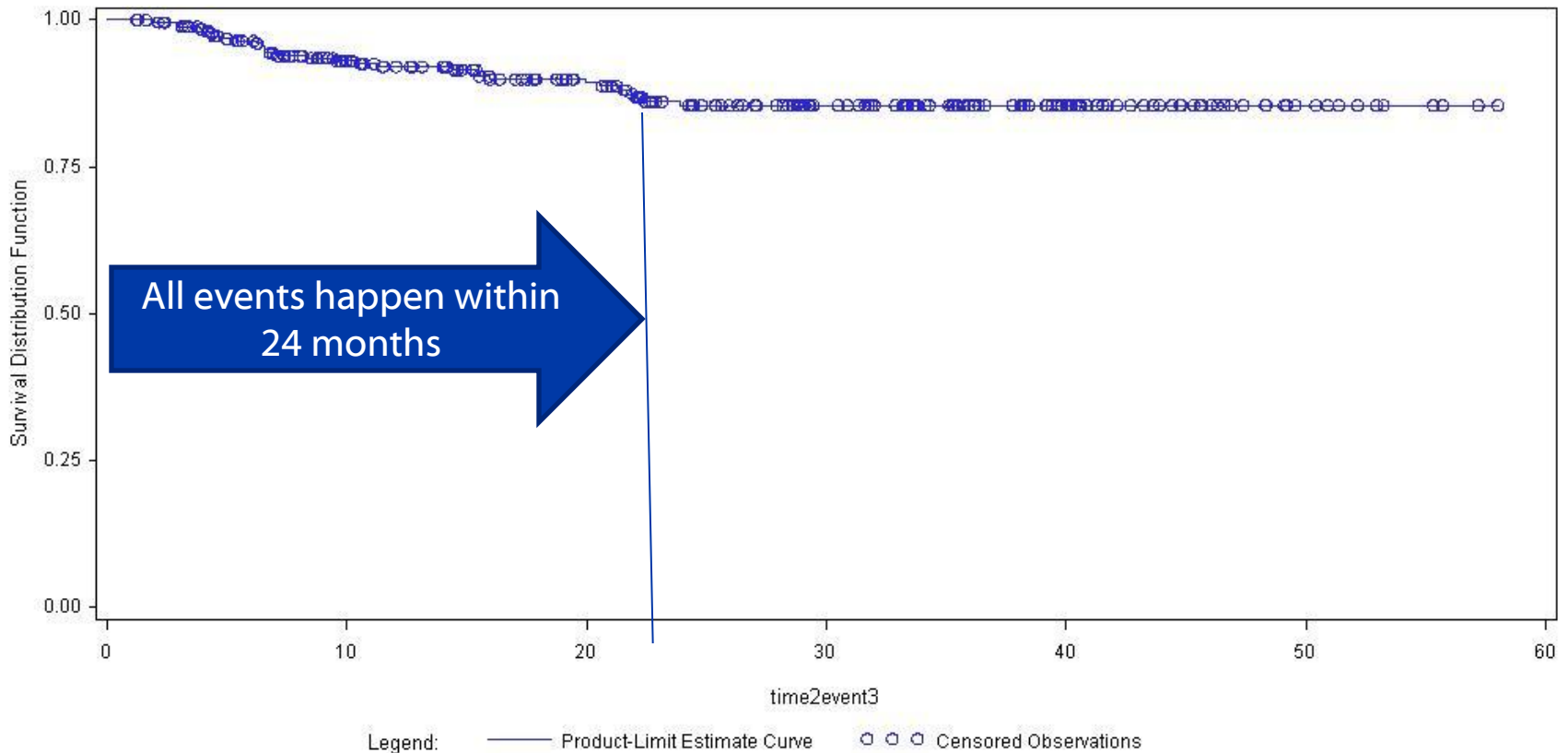
- ❑ Identified by SaTScan (2006-2010)**
- ❑ Clusters ≥ 3 cases with same genotype**
- ❑ Annual county genotyping coverage $\geq 75\%$**
- ❑ No matching cases in preceding 24 months**
 - Identify new (i.e., incident) clusters**

How we defined our outcome: “outbreak”

From beginning (January 2006) to end (Dec 2010) of observation period:

- Grew from 3 to ≥ 6 cases
- Within 24 months
- Confirmed by local public health officials as an outbreak

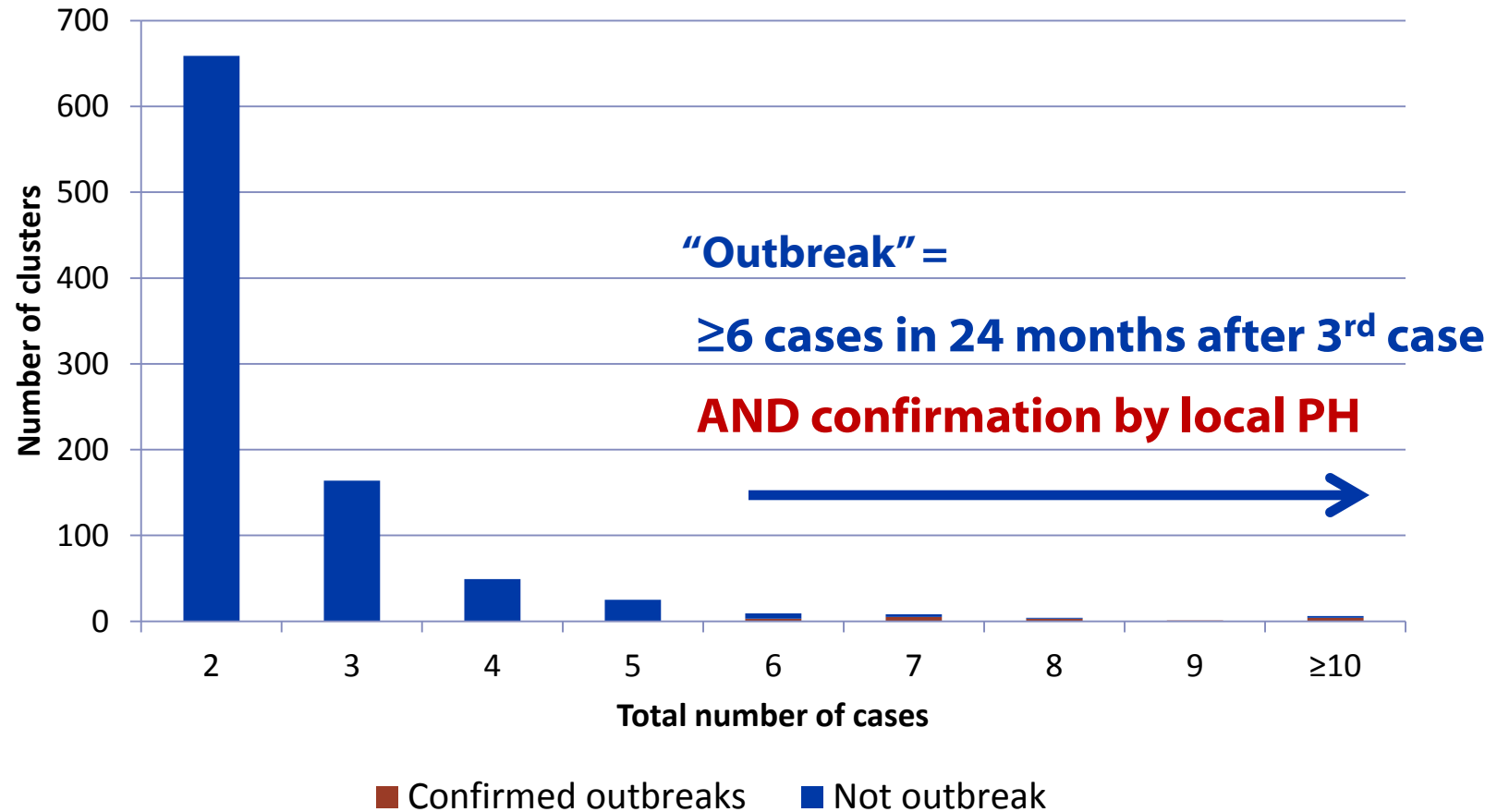
Time to reach 6 cases, N=146



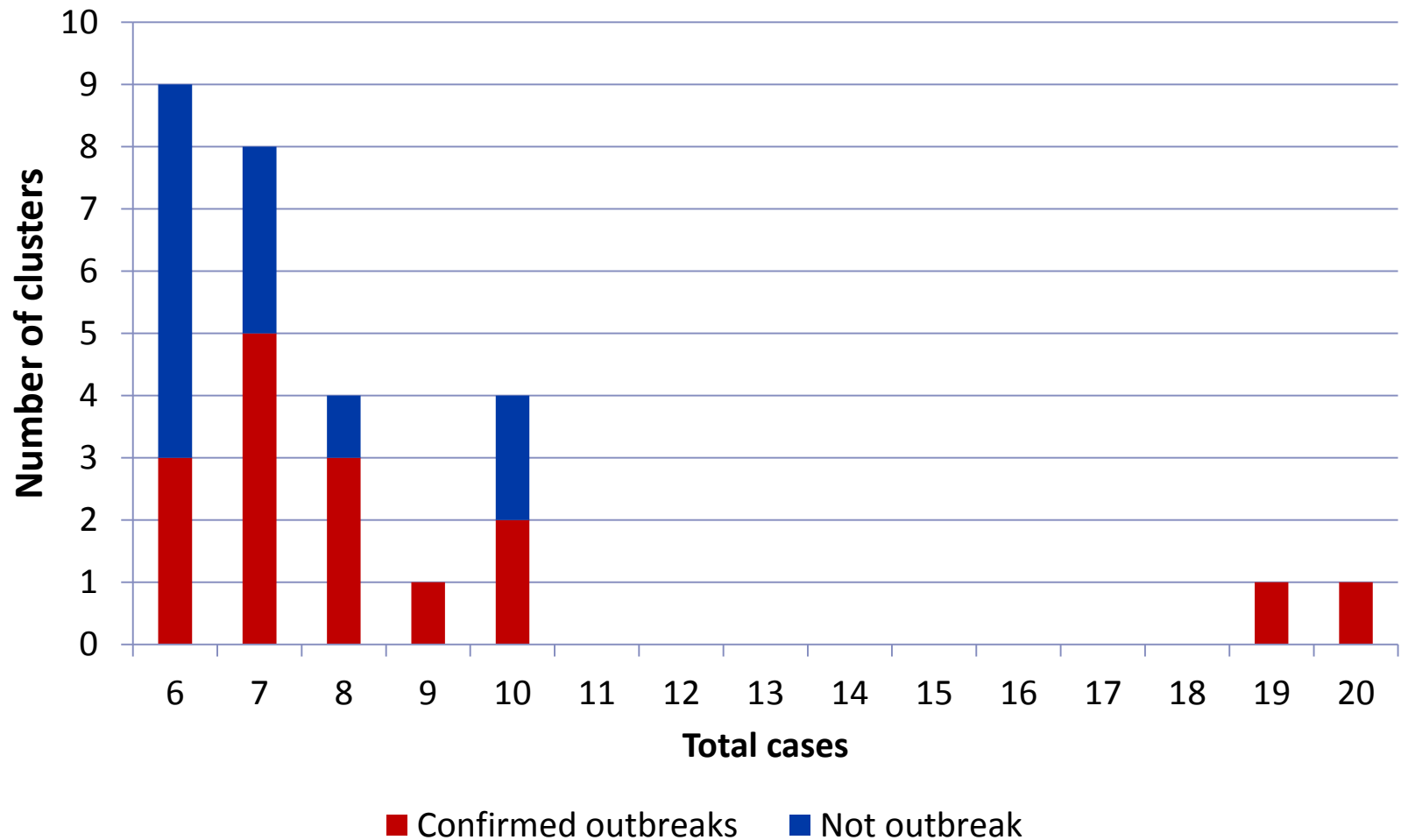
Entry into cohort: 3rd case

Event: 6th case in an outbreak cluster of $n \geq 6$, else censored at end of timeline (Dec 2010)

Distribution of clusters by final cluster size, stratified by outcome, N=925



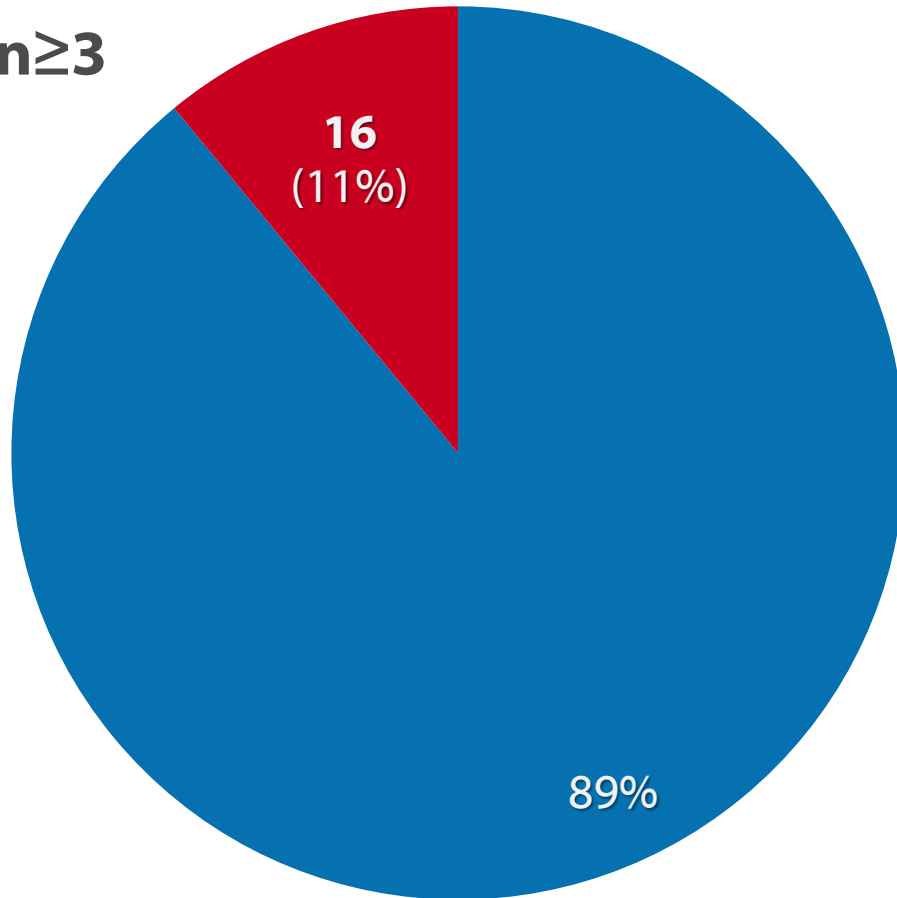
Distribution of clusters by final cluster size, stratified by outcome, $n \geq 6$



Outcomes

□ 146 clusters of size $n \geq 3$

- Not outbreaks
- Outbreaks



Characteristics of 146 TB clusters and associated risk factors of cluster becoming an outbreak case

Cluster characteristic	Outbreak clusters (n=16) with characteristic		Relative risk in predicting outbreaks	
	n	%	RR	95% CI
Homeless or excess alcohol use or illicit drug use or incarceration at diagnosis	15	22.4	17.7	2.4, 130.4
Homeless or alcohol or drug use	14	21.9	9.0	2.1, 38.0
Drug use	12	25.5	6.3	2.2, 18.6
Alcohol use	11	25.0	5.1	1.9, 13.8
Homeless or alcohol use	11	22.0	4.2	1.6, 11.5
Homeless and alcohol use	6	30.0	3.8	1.5, 9.3
Incarceration	5	29.4	3.4	1.4, 8.7
Homeless	6	23.1	2.8	1.1, 6.9

Characteristics of 146 TB clusters and associated risk factors of cluster becoming an outbreak case

Cluster characteristic	Outbreak clusters (n=16) with characteristic		Relative risk in predicting outbreaks	
	n	%	RR	95% CI
T1 to T3 <5.3 months	9	21.4	3.2	1.3, 8.0
T1 to T2 <4.4 months	12	16.9	3.2	1.1, 9.4
T2 to T3 <0.9 months	7	21.2	2.7	1.1, 6.6
Sig. LLR at 3 rd case	13	16.3	3.5	1.0, 11.6

Decision Tree analysis using JMP

- ❑ **SAS Predictive modeling software**
- ❑ **Recursive partitioning function**
 - “enables users to systematically analyze large data sets to discover unsuspected or unknown relationships.
 - finds a set of cuts or groupings of X values that best predict an outcome by exhaustively searching all possible cuts or groupings, recursively forming a tree of decision rules until the desired fit is reached.”

Algorithm for predicting outbreaks at 3rd case

146 new clusters of 3 patients;
16 (11.0%) became outbreaks

67 with ≥ 1 of 3 patients who reported excess alcohol use, homelessness, illicit drug use, or incarceration at diagnosis;
15 (22.4%) became outbreaks

Low-Risk Clusters
79 with 0 of 3 patients who reported excess alcohol use, homelessness, illicit drug use, or incarceration at diagnosis;
1 (1.3%) became an outbreak

High-Risk Clusters
17 had 3rd patient diagnosed < 5.3 months after 1st patient;
9 (52.9%) became outbreaks

Medium-Risk Clusters
50 had 3rd patient diagnosed ≥ 5.3 months after 1st patient;
6 (12.0%) became outbreaks

Predicting outbreaks at 3rd case

146 new clusters of 3 patients
11% became outbreaks

High Risk
53% became outbreaks
**Targets for
intervention**

Medium risk
12% became outbreaks
Watch list

Low risk
1% became outbreaks
Back burner

Limitations

- ❑ **Surveillance data**
- ❑ **Observation period of only 5 years**
- ❑ **75% genotyping coverage**
(We now have >90% coverage)
- ❑ **Based on 12-locus MIRU**
(We now have 24-locus MIRU data)

Conclusion

- ❑ **With this methodology, we are able to predict half of clusters that would become outbreaks**
- ❑ **This provides an approach to prioritizing clusters for early intervention**

Acknowledgements

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