

# Pooled testing with penalized regression models

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  - Efficient

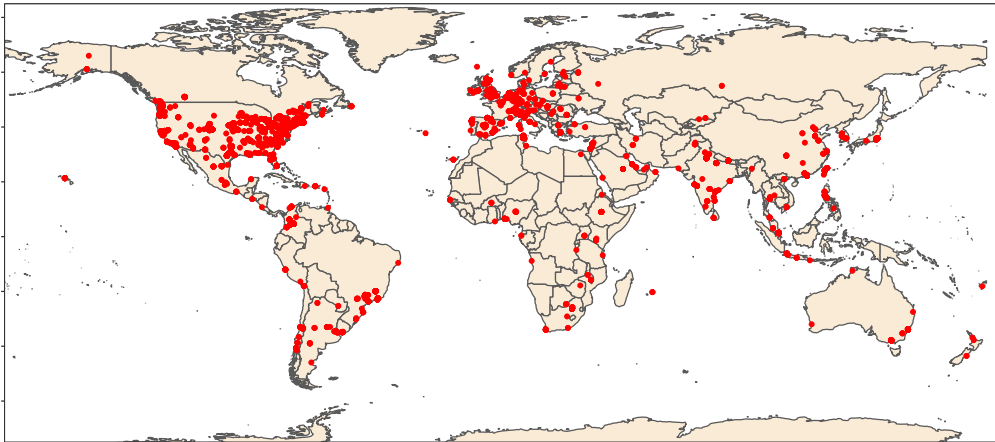
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  - Decrease number of tests, increase testing capacity

- Widely used during pandemic
  - A Shiny App for Pooled Testing
  - 91 countries during 6 months of 2020



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- Department of Health, Disability and Ageing: “Revised testing framework for COVID-19 in Australia, March 2022”
  - Microbiological Diagnostic Unit Public Health Lab at U. of Melbourne (Chong et al. 2020)
- Widely used elsewhere: Blood donations, sexually transmitted infections, congenital infections, animal infections, food safety surveillance, computer networks assessments, flower infection levels

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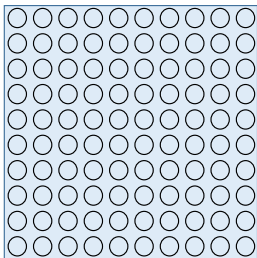


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  - Use viral load responses rather than binary (positive/negative) responses
  - Use linear model to predict positive/negative

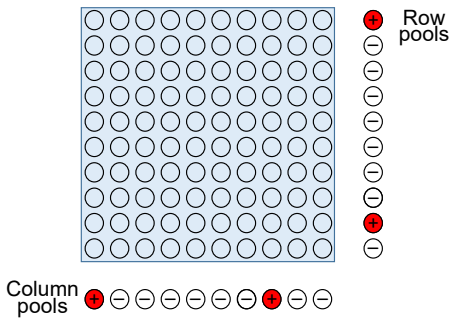
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- Purpose: Examine use of viral load response and linear model prediction with “array testing” algorithm

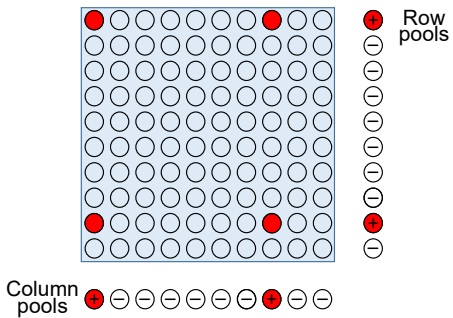
## ● Array testing



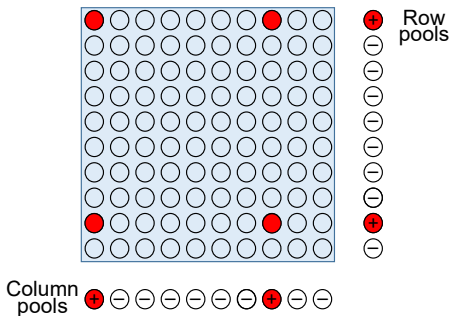
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- **Key aspect:** Test in multiple pools (groups) during first stage to reduce the number of retests in a second stage

- A  $3 \times 3$  array

	Column1	Column 2	Column 3
Row 1	1	2	3
Row 2	4	5	6
Row 3	7	8	9



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## ● Alternative form

Pools	Specimens									Regular Array
	1	2	3	4	5	6	7	8	9	
1	1	1	1	0	0	0	0	0	0	Row 1
2	0	0	0	1	1	1	0	0	0	Row 2
3	0	0	0	0	0	0	1	1	1	Row 3
4	1	0	0	1	0	0	1	0	0	Col. 1
5	0	1	0	0	1	0	0	1	0	Col. 2
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- Pooling matrix:  $X_{6 \times 9}$  is a matrix of 0's and 1's

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  - Penalized regression model: non-negative LASSO

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  - #2: Same as #1 but
    - Retest specimens in a second stage for estimates in an indeterminate range
    - Indeterminate range:  $0 < \hat{\beta}_i < c$

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- Estimate efficiency and PPA
  - No closed form expressions for linear model-based algorithms
  - Use Monte Carlo simulation

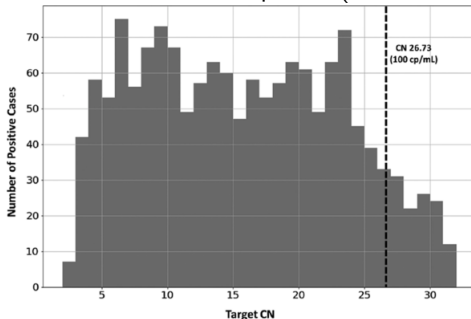


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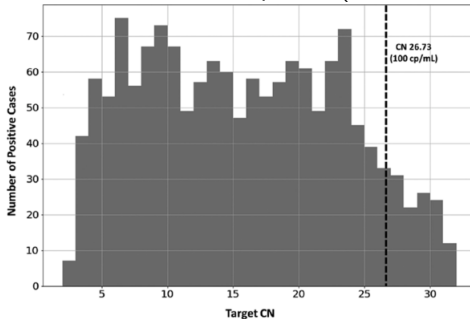
## ● Monte Carlo simulation summary

- Simulate individual positive/negative status with Bernoulli( $p$ ),  $p$  is infection prevalence
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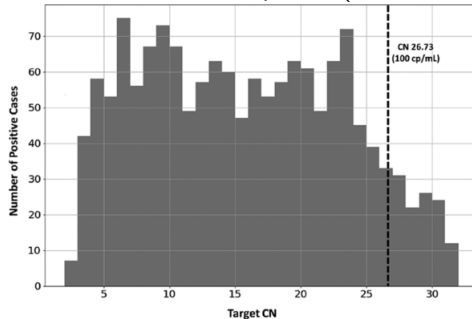
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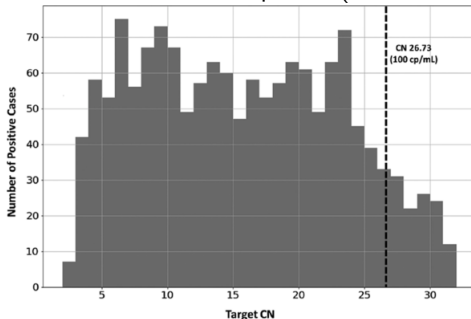
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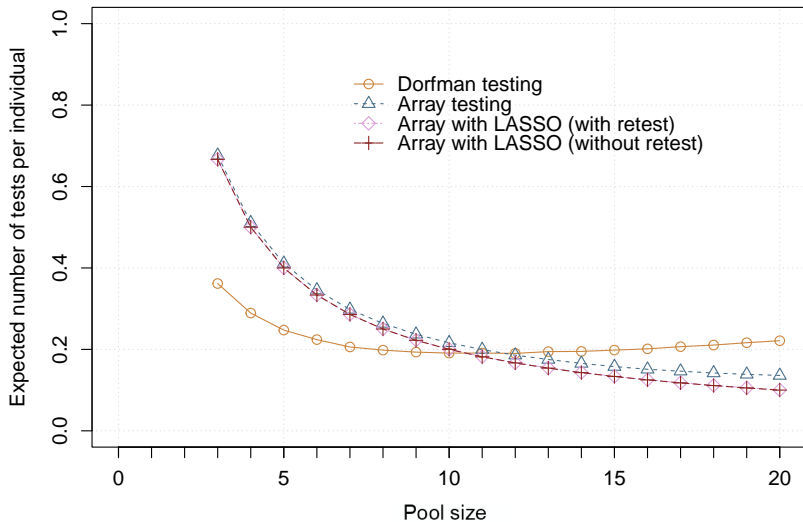
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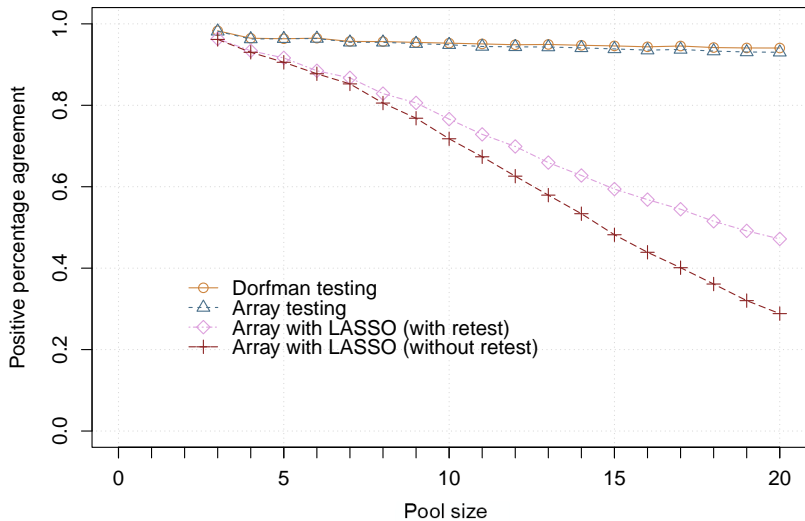
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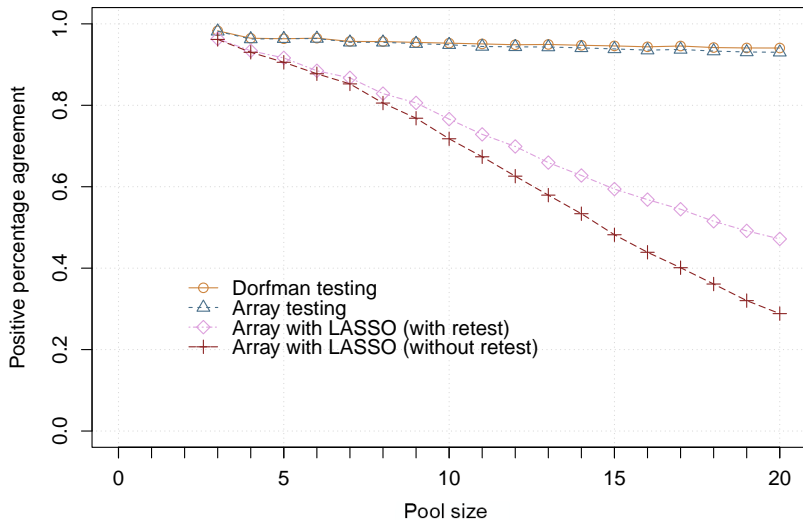


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- One setting:  $p = 0.01$ , pool sizes 3 to 20









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- Incorporate statistical inference -  $H_0 : \beta_i = 0$  vs.  $H_a : \beta_i > 0$

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