

**\*CRIG** 



# COVID-19 sample pooling: to pool or not to pool?

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**Test everybody!** 40 people, 40 tests

Many tests... to find one patient...



#### Let's make pools!

- 1. Test each pool as a whole
- 2. Test samples from positive pool individually

#### 40 people, 4 + 10 = **14 tests**

### Why pooling?

#### 1. Massive population screening

#### 2. Lower cost

3. Quicker results for more people

- 1. One-time pooling (1D)
- 2. 2D-pooling
- 3. Sequential pooling
- 4. P-BEST or Tapestry

#### 1. One-time pooling (1D)

- 1. Make pools and test
- 2. Positive pools tested individually
- 2. 2D-pooling
- 3. Sequential pooling
- 4. P-BEST or Tapestry



- 1. One-time pooling (1D)
- 2. 2D-pooling
  - 1. Arrange in grid
  - 2. Make pools over columns and rows
  - 3. Test pools
  - 4. Remove negative rows and columns
  - 5. Test individual samples that remain
- 3. Sequential pooling
- 4. P-BEST or Tapestry





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  - Go back to the same samples over and over
- 4. P-BEST and Tapestry
  - Complex pipetting regimes

## Not practical

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- 2. 2D-pooling
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  - · Go back to the same samples over and over

#### 4. P-BEST and Tapestry

Complex pipetting regimes

## Not practical

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### Simulation approach

- 1. Use viral load proxy (Cq value) instead of binary absent/present (negative/positive)
  - Dilution during pooling: *false-negatives*

#### 2. Screening

- 1. High number of samples
- 2. Low number of infected/positive people
- 3. Practical pools
  - Think about microtiter plates (4, 8, 12, 16, 24)
- 4. Use real life data

#### **Real-life data**

- Real Cq values from national testing platform
- Full 96-well RNA plates with less than 10 positives and good controls



Cq value

Mimic low prevalence screening

#### **Experimental set-up**

- 1. No. of samples: 100 000 samples
- 2. Prevalence: 0.01% to 10%
- 3. Strategy:
  - 1D-pooling: 4, 8, 12, 16, 24 samples
  - 2D-pooling: 8x12, 12x16, 16x24 matrix
- 4. Replicate simulations: 5



Efficiency increase

# of tests without pooling

*# of tests with pooling* 

• Sensitivity

*# of true positives* 

*# of true positives + # of false negatives* 





average efficiency increase



prevalence (%)



prevalence (%)



# 'Rescue'

Low viral load samples, that in a pool would test <u>negative due to dilution</u>, can <u>test positive when a high viral load sample is present</u> in the same pool



prevalence (%)





prevalence (%)

efficiency







sensitivity

#### Take-aways

- 1. At high prevalence, 2D is more efficient
- 2. At low prevalence, 1D is more efficient
- 3. Large pools have **better efficiency**
- 4. Large pools have more false-negatives

#### Evaluation of efficiency and sensitivity of 1D and 2D sample pooling strategies for diagnostic screening purposes Jasper Verwilt, Pieter Mestdagh, Jo Vandesompele



medRχiv https://www.medrxiv.org/content/10.1101/2020.07.17.20152702v2



- Jo Vandesompele
- Pieter Mestdagh
- Alexander Reinartz
- OncoRNALab



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