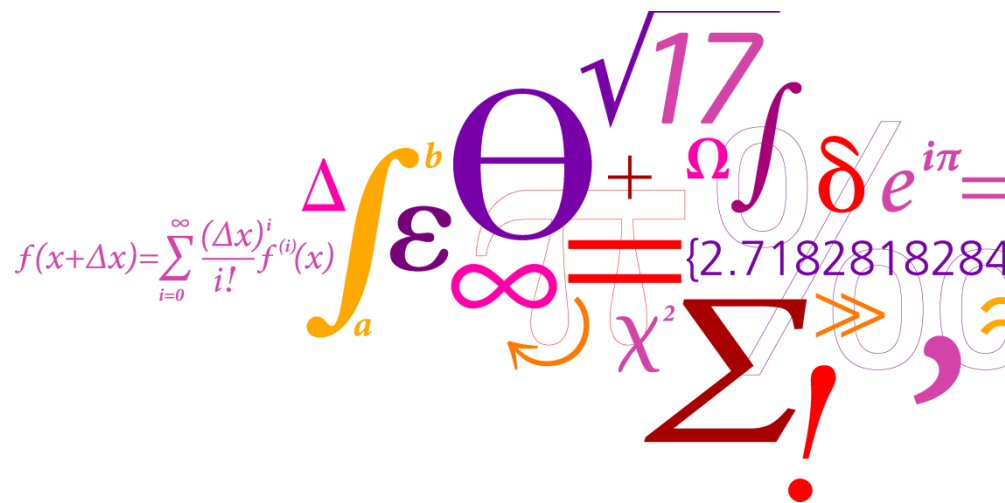


# Group testing – pool sizes, effects and group sensitivity

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# Why Group testing

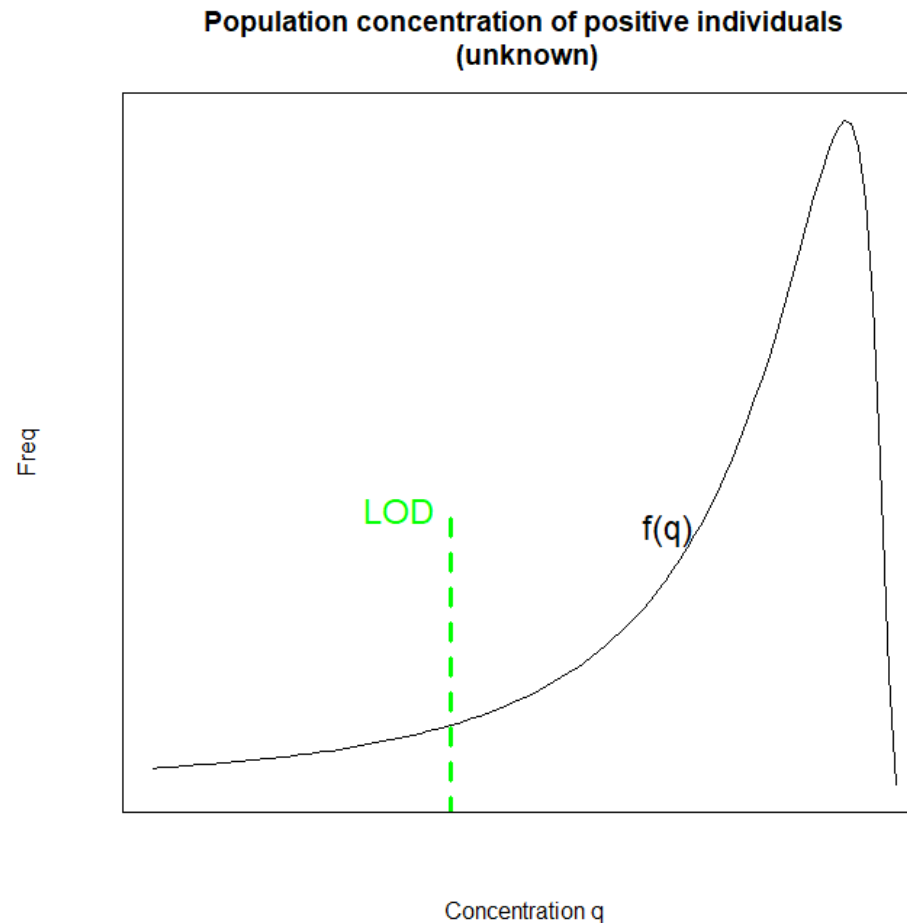
- Fewer tests save:

**Equipment**

**Time**

# Sars-Cov-2 testing

- RT PCR, result is a score in copies/ $\mu$ L.



# Imperfect tests

- Even if a person is infected, the sample may not contain virus ( throat swab in the case of a lung infection)
- Even if a sample from an individual contains virus, the test may not detects it. Reasons:
  - The concentration is below the Limit of Detection (LOD);
  - The test fail to accurately measure the concentration.

The probability of a positive sample given infection is expressed as the sensitivity  $\lambda$ .

# Probabilistic approach to tests

Assume that the test is being conducted for a disease with prevalence  $p$ . The test has a sensitivity of  $\lambda$ , ie.

$$P(\textit{Test positive} | \textit{Disease positive}) = P(T + | D +) = \lambda$$

For simplicity, perfect specificity is assumed, ie.

$$P(\textit{Test negative} | \textit{Disease negative}) = P(T - | D -) = 1$$

# Probabilistic approach to tests

The **test result**  $\mathbb{X}$  depends on both the **concentration**  $q$  (with density  $f$  for positive individuals), and on **measurement uncertainty**.

Will assume a multiplicative conditional model so that

$$\log(\mathbb{X}|q) \sim N(\log(q), \sigma^2)$$

Ie. Conditionally on  $q$ ,  $\mathbb{X}$  is log-normal:

$$\log(\mathbb{X}) = \log(q) + \sigma\varepsilon$$

**concentration**

**measurement  
uncertainty**

(this assumption only plays a part for the last part of the talk)

The test is **positive** if  $\mathbb{X} > LOD$  (another value could apply)

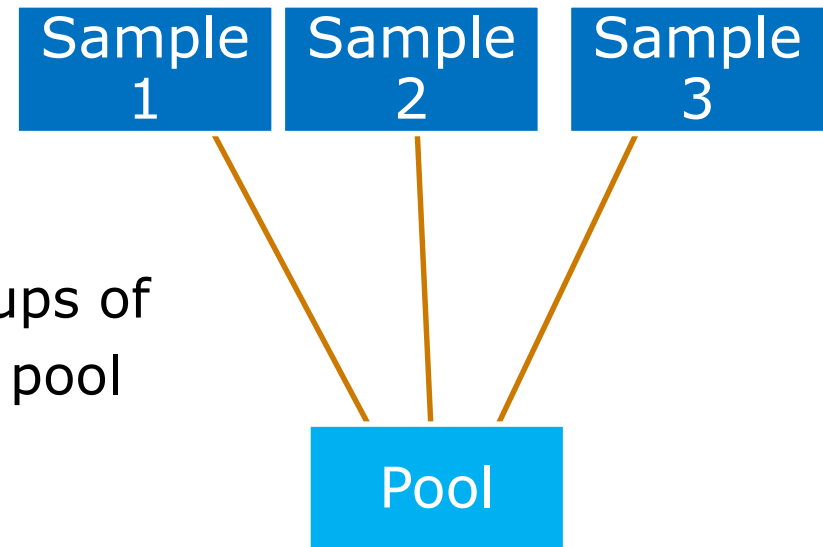
# Pool tests

- Simple pooling scheme:

- **Divide** the subjects into groups of size  $k$  for some integer, and pool their samples.

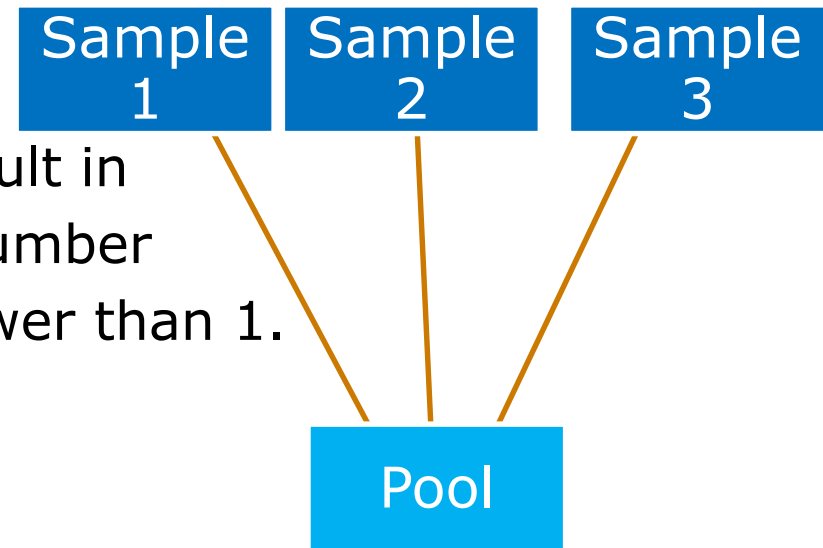
- **Test** the pooled sample;

- If the pooled sample is positive, **retest** every individual in the group.



# Pooled tests

- Well chosen group sizes will result in a lesser **effort**; the expected number of tests per individual will be lower than 1.
- The effort is ( $k > 1$ )



$$\lambda_k(1 - (1 - p)^k) + \frac{1}{k}$$

Ex.  $k = 1$ : Effort is 1 ☹ .

Ex.  $k = n$ : Effort may be  $> 1$ .

Somewhere in between is better...



# Minimizing the effort

For a fixed sensitivity, the group size with minimum effort can be determined analytically as

$$k = 2 \operatorname{Argmin} \left( W_0 \left( -\sqrt{-\frac{\log(1-p)}{4\lambda}} \right), \left( W_{-1} \left( -\sqrt{-\frac{\log(1-p)}{4\lambda}} \right) \right) \right) / \log(1-p)$$

Where  $W_0$ ,  $W_{-1}$  are the two branches of the Lambert function, and 'Argmin' chooses the one with the lowest effort (for realistic values of  $p$  it will be  $W_0$ ).

	$p = 0.1$		$p = 0.01$		$p = 0.001$		$p = 0.0001$		$p = 0.00001$	
	$k$	Effort	$k$	Effort	$k$	Effort	$k$	Effort	$k$	Effort
$\lambda = 1$	4	0.59	11	0.20	32	0.06	101	0.02	317	0.006
$\lambda = 0.9$	4	0.56	11	0.19	34	0.06	106	0.02	334	0.006
$\lambda = 0.8$	4	0.53	12	0.17	36	0.06	112	0.02	354	0.006
$\lambda = 0.7$	5	0.49	13	0.16	39	0.05	120	0.02	379	0.005

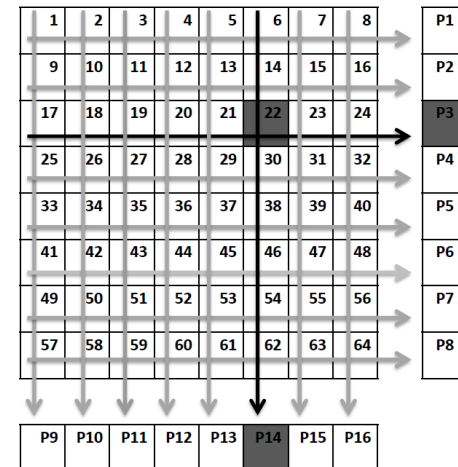
# What if the the pool test is positive?

- The same sample appears in multiple pools: Results are stochastically **dependent** through the concentration  $q$ ;

Consider

$$P(\mathbb{X}_2 \text{ positive}, \mathbb{X}_{pool} \text{ positive}) = \lambda_k P(\mathbb{X}_2 \text{ positive} | \mathbb{X}_{pool} \text{ positive})$$

- If  $\mathbb{X}_2$  is a single test result, it is the probability that **simple pooling** catches a case;
- If  $\mathbb{X}_2$  is a pooled test containing the same (case) individual, it is the probability that Kaare Græsbøls's **2nd order non-hierachical testing** catches a case.



# What if the the pool test is positive?

## - Initial thoughts

- With knowledge on  $f$  and  $\sigma^2$ , one may derive (approximate) expressions for the probability that a case will result in a detection.
- Disregard two positive samples in the pool (low prevalence), assume that the sample **dilutes the concentration**  $k$ -fold and consider the event

$$\{\mathbb{X}_{pool} > LOD\}; \text{ ie. } \{\log(\mathbb{X}_{pool}) = \log(q/k) + \sigma\varepsilon >$$

# What if the the pool test is positive? - Initial thoughts – loss of sensitivity

$$\{\mathbb{X}_{pool} > LOD\}; \text{ ie. } \{\log(\mathbb{X}_{pool}) = \log(q/k) + \sigma\varepsilon >$$

# Conditional probability of a single test being positive

Thus:

$$P(\mathbb{X}_{single} \text{ positive} | \mathbb{X}_{pool} \text{ positive}) \approx$$

$$P(\mathbb{X}_{single} \text{ positive} | q > kLOD e^{-1.64\sigma}) =$$

$$\int_{\log(kLOD) - 1.64\sigma}^{\infty} \int_{LOD}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x - \log(q))^2} f(q) dx dq$$

- For known values of  $f$  and  $\sigma^2$  this integral can be calculated, and provides the approximate probability that an individual is tested positive, given the pooled sample is positive.
- Together with  $\lambda_k$ , this is a measure of the loss in sensitivity.

# Research

- More thorough expressions of conditionally positive tests.
- Detect analytical expressions for optimal pool sizes with varying sensitivity;
- Derive optimal pool size analytically under more advanced pooling schemes than simple pooling;
- **How to get the parameter values?**
  - Need  $f$  and  $\sigma^2$ ; can be derived from (at least) double measurements on the same sample. Will be dependent on lab technician, kit, lab, etc. Should be found from a random effects model. Looking into possibilities with Johan, Kaare and others.