# Estimation of the effective reproduction number of a measles outbreak in Guinea, 2017

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Objective

Assess the association between outbreak response vaccination (ORV) campaigns and the effective reproduction number  $(R_t)$ of measles during a nationwide outbreak in Guinea.

Guinea



Median age: 2 years (96% < 15 years)



## Background

The **measles virus** is spread through droplets (e.g., via coughing) but can also become airborne.

Measles is a vaccine-preventable disease. 2016 vaccination coverage in Guinea: 48% (target: 95%)

A nationwide outbreak began early January 2017 following disruptions to vaccination during the 2014–2016 Ebola epidemic (1)

**Outbreak response vaccination (ORV)** campaigns were implemented to reduce transmission, morbidity, and mortality (2):

#### Guinea measles ORV campaigns in 2017

Campaign	Campaign dates	# of health districts	Target age (months)	Number vaccinated
1	Mar 13–19	1	6–119	148,344
2	Apr 9–17	5	6–119	662,733
3	Apr 25–May 1	22	6–59	1,315,918



#### Effective reproduction number based on ORV campaign timing

Timing of analysis	Measles effective reproduction number (95% CI)		
Start of ORV campaign	Four parameters	<b>Two parameters</b>	
Before ORV	1.64 (1.58–1.70)	1.64 (1.58–1.71)	
ORV campaign 1	0.71 (0.67–0.75)	0.75 (0.73–0.78)	
ORV campaign 2	0.89 (0.82–0.96)	$\downarrow$	
ORV campaign 3	0.71 (0.67–0.75)	$\downarrow$	
AIC	2579	2654	

#### Measles outbreak cases and effective reproduction number, 2017



ORV campaign success is quantifiable using the effective reproduction **number**  $R_t$ , which estimates the average number of secondary cases produced by an infectious individual within a partially immune population.

## Methods

We estimated R<sub>f</sub> for all measles cases with illness onset in calendar year 2017 using a pulse time-dependent step function...

$$R_t = egin{cases} R_1, & ext{for } t < t_0 \ R_2, & ext{for } t_0 \leq t < t_1 \ R_3, & ext{for } t_1 \leq t < t_2 \ R_4, & otherwise \end{cases}$$

...and generalized renewal equation...

- $E(i_t) = R_t \sum_{i=1}^{t-i} i_{t-s} g_s$
- $E(i_t)$  = estimated incidence at time t
- = reproduction number at time *t*
- = incident cases at time t
- = probability mass function of the  $g_{s}$ measles virus generation time using a gamma distribution

#### Discussion

- The model captured the peaks of the outbreak well, but does not account for spatial heterogeneity of cases.
- Model fit was better when cutpoints were set to the beginning of each ORV campaign (AIC=2579) rather than the end of the buffer period (2) weeks after the beginning of a campaign, AIC=3204).



...and minimized the negative log likelihood.

- 95% confidence intervals were computed using profile likelihood.
- Akaike information criterion (AIC) values and visual model fit were used to select the final model.

### References

- (1) Takahashi S, et al. The growing risk from measles and other childhood infections in the wake of Ebola. Science. 2016;347(6227):1240–2.
- (2) Minetti A, et al. Measles outbreak response immunization is contextspecific: Insight from the recent experience of Médecins Sans Frontières. PLoS Medicine. 2013;10(11):1-4.

#### Conclusions

- ORV implementation coincided with the decrease in  $R_t$  to < 1.
- Calculation of  $R_t$  from case count data using basic modeling methods can help decision makers and those in field understand progress of an outbreak and make decisions about initiating further control measures.

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