## S2 Text

Proofs on Screening Requirements for Eradication. For sake of conciseness we omit the subscript  $v$  in this section. Before we state our proofs, let us make some observations about the expected prevalence level immediately after the  $n^{th}$  screening round:

$$
f(S_n^+) = \frac{K}{1 + A_n} \tag{1}
$$

Let us specifically focus on the parameter  $A_n$ . Note that a larger value of  $A_n$  implies a lower prevalence level after the screening round and vice versa. Recall that the impact of a screening round is estimated as (see main text):

$$
f_v(S_{vn}^+) = (1 - p_{vn} \cdot s) f_v(S_{vn}^-),
$$

This implies for any  $n \geq 1$ :

$$
A_n = \frac{K}{f(S_n^+)} - 1\tag{2}
$$

$$
=\frac{K}{(1-p\cdot s)\cdot f(S_n^-)}-1\tag{3}
$$

$$
=\frac{K}{(1-p\cdot s)\cdot\frac{K}{1+A_{n-1}\cdot e^{-\kappa\cdot\tau}}}-1\tag{4}
$$

$$
= \alpha \cdot A_{n-1} + \beta \tag{5}
$$

<span id="page-0-1"></span>
$$
= \alpha^n \cdot A_0 + \sum_{i=1}^n \alpha^{n-i} \cdot \beta \tag{6}
$$

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$$
= \begin{cases} \alpha^n \cdot A_0 + \beta \frac{\alpha^n - 1}{\alpha - 1} & \text{if } \alpha \neq 1 \\ A_0 + n \cdot \beta & \text{if } \alpha = 1 \end{cases}
$$
 (7)

Here,  $\alpha = \frac{1}{1-p \cdot s} e^{-\kappa \cdot \tau}$ ,  $\beta = \frac{p \cdot s}{1-p \cdot s}$ , and  $A_0 = \frac{K}{f(0)} - 1$ . Note that  $\alpha$ ,  $\beta$ , and  $A_0$  are strictly positive. These observations and definitions allow us to prove the following results.

<span id="page-0-2"></span>**Lemma 1.** If  $\tau \leq \frac{-\log(1-p \cdot s)}{r}$  $\frac{(1-p \cdot s)}{\kappa}$ , then  $\lim_{n \to \infty} A_n = \infty$  and  $A_n$  is discretely convex and increasing in n.

*Proof.* First, rewriting yields that the condition  $\tau \leq \frac{-\log(1-p\cdot s)}{r}$  $\frac{1-p \cdot s}{\kappa}$  is equivalent to the condition  $\alpha = \frac{1}{1-p\cdot s}e^{-\kappa \cdot \tau} \geq 1$ . The result now follows immediately from Eq. [\(7\)](#page-0-0) and by the observations that  $A_0 > 0$  and  $\beta > 0$ . If  $\alpha > 1$  these imply that:

$$
\lim_{n \to \infty} A_n > \lim_{n \to \infty} \alpha^n A_0 = \infty
$$
\n(8)

Furthermore, if  $\alpha = 1$ , it holds that:

$$
\lim_{n \to \infty} A_n = \lim_{n \to \infty} A_0 + n \cdot \beta = \infty
$$
\n(9)

The discrete convexity of  $A_n$  is trivial for the case that  $\alpha = 1$ , since  $A_n$  is linear in n in this case. If  $\alpha > 1$ , the convexity follows from the fact that every function having the form  $c \cdot \alpha^n + d$  is convex in n for  $c \ge 0$  and  $\alpha \ge 0$ . Finally, the fact that  $A_n$  is increasing in  $n$  follows directly from Eq  $(6)$ .

 $\Box$ 

<span id="page-1-0"></span>**Lemma 2.** If  $\tau > \frac{-\log(1-p\cdot s)}{\kappa}$ , then then the sequence  $\{A_n\}$  converges monotonically to  $A=\frac{\beta}{1-\alpha}$ .

*Proof.* First, rewriting yields that the condition  $\tau > \frac{-\log(1-p \cdot s)}{\kappa}$  is equivalent to the condition  $\alpha = \frac{1}{1-p \cdot s} e^{-\kappa \cdot \tau} < 1$ . The fact that the sequence converges to A now follows immediately by taking the limit in Eq. [\(7\)](#page-0-0):

$$
\lim_{n \to \infty} A_n = \lim_{n \to \infty} \alpha^n \cdot A_0 + \beta \frac{\alpha^n - 1}{\alpha - 1} \tag{10}
$$

$$
=\frac{\beta}{1-\alpha}=A\tag{11}
$$

To prove that  $A_n$  converges monotonically to  $A$ , we distinguishing two cases:  $A_{n-1} < A$  and  $A_{n-1} > A$ , (it is obvious for  $A_{n-1} = A$ ). First, consider the case that  $A_{n-1} < A$ . We show that  $A_n < A$ , and that  $A_n - A_{n-1} > 0$ , implying monotonicity. The first claim is implied by Eq. [\(6\)](#page-0-1):

$$
A_n = \alpha \cdot A_{n-1} + \beta \tag{12}
$$

$$
\langle \alpha \cdot \frac{\beta}{1-\alpha} + \beta \tag{13}
$$

$$
=\frac{\beta}{1-\alpha}=A\tag{14}
$$

To prove the second claim we make use of the given that  $\alpha < 1$ :

$$
A_n - A_{n-1} = \alpha \cdot A_{n-1} + \beta - A_{n-1}
$$
\n(15)

$$
= (\alpha - 1) A_{n-1} + \beta \tag{16}
$$

$$
> (\alpha - 1) \frac{\beta}{1 - \alpha} + \beta = 0 \tag{17}
$$

Next, consider the case that  $A_{n-1} > A$ . Using on exactly the same reasoning as the one presented above, it follows that  $A_n > A$  as well, and that  $A_n - A_{n-1} < 0$ .

This completes the proof.

Now we have established the (long term) behavior of the parameter  $A_n$ , let us consider the implications for the (long term) prevalence level. Recall that  $\bar{f}_{n,n+1}$ denotes the expected average prevalence level faced between screening rounds  $n$  and  $n + 1$ , which is calculated as:

$$
\bar{f}_{n,n+1} = \frac{1}{\tau} \int_0^{\tau} \frac{K}{1 + A_n \cdot e^{-\kappa \cdot t}} dt
$$
\n(18)

$$
= \frac{K}{\kappa \cdot \tau} \log \left( \frac{A_n + e^{\kappa \cdot \tau}}{A_n + 1} \right) \tag{19}
$$

Using this definition, we now prove the desired results:

**Proposition 1.** If  $\tau \leq \frac{-\log(1-p \cdot s)}{s}$  $\frac{(1-p\cdot s)}{\kappa}$ , then  $\lim_{n\to\infty} \bar{f}_{n,n+1} = 0$ .

*Proof.* From Lemma [1](#page-0-2) we know that  $\lim_{n\to\infty} A_n = \infty$ . Hence, we derive that:

 $\Box$ 

<span id="page-2-0"></span>
$$
\lim_{n \to \infty} \bar{f}_{n,n+1} = \lim_{n \to \infty} \frac{K}{\kappa \cdot \tau} \log \left( \frac{A_n + e^{\kappa \cdot \tau}}{A_n + 1} \right) \tag{20}
$$

$$
=\frac{K}{\kappa \cdot \tau} \log\left(1\right) = 0\tag{21}
$$

 $\Box$ 

**Proposition 2.** If  $\tau > \frac{-\log(1-p \cdot s)}{\kappa}$ , then  $\lim_{n \to \infty} \bar{f}_{n,n+1} = K\left(\frac{\log(1-p \cdot s)}{\kappa \cdot \tau} + 1\right)$ .

*Proof.* From Lemma [2](#page-1-0) we know that  $\lim_{n\to\infty} A_n = A = \frac{\beta}{1-\alpha}$ . Hence, we derive that:

$$
\lim_{n \to \infty} \bar{f}_{n,n+1} = \lim_{n \to \infty} \frac{K}{\kappa \cdot \tau} \log \left( \frac{A_n + e^{\kappa \cdot \tau}}{A_n + 1} \right) \tag{22}
$$

$$
= \frac{K}{\kappa \cdot \tau} \log \left( (1 - p \cdot s) e^{\kappa \cdot \tau} \right) \tag{23}
$$

$$
= K\left(\frac{\log\left(1 - p \cdot s\right)}{\kappa \cdot \tau} + 1\right) \tag{24}
$$

 $\Box$