OPEN LETTER

Insights from quantitative and mathematical modelling on the proposed 2030 goal for gambiense human African trypanosomiasis (gHAT) [version 2; peer review: 2 approved]

NTD Modelling Consortium Discussion Group on Gambiense Human African Trypanosomiasis

Abstract
Gambiense human African trypanosomiasis (gHAT) is a parasitic, vector-borne neglected tropical disease that has historically affected populations across West and Central Africa and can result in death if untreated. Following from the success of recent intervention programmes against gHAT, the World Health Organization (WHO) has defined a 2030 goal of global elimination of transmission (EOT). The key proposed indicator to measure achievement of the goal is zero reported cases. Results of previous mathematical modelling and quantitative analyses are brought together to explore both the implications of the proposed indicator and the feasibility of achieving the WHO goal.

Whilst the indicator of zero case reporting is clear and measurable, it is an imperfect proxy for EOT and could arise either before or after EOT is achieved. Lagging reporting of infection and imperfect diagnostic specificity could result in case reporting after EOT, whereas the converse could be true due to underreporting, lack of coverage, and cryptic human and animal reservoirs. At the village-scale, the WHO recommendation of continuing active screening until there are three years of zero cases yields a high probability of local EOT, but extrapolating this result to larger spatial scales is complex. Predictive modelling of gHAT has consistently found that EOT by 2030 is unlikely across key endemic regions if current medical-only strategies are not bolstered by improved coverage, reduced time to detection and/or complementary vector control. Unfortunately, projected costs for strategies expected to meet EOT are high in the short term and strategies that are cost-effective in reducing burden are unlikely to result in EOT by 2030. Future modelling work should aim to provide predictions while taking into account uncertainties in stochastic dynamics and infection reservoirs, as well as assessment of multiple spatial scales, reactive strategies, and measurable proxies of EOT.

Open Peer Review

Approval Status

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2. Mario Recker, University of Exeter, Penryn, UK

Any reports and responses or comments on the article can be found at the end of the article.
Keywords
 gambiense human African trypanosomiasis (gHAT), sleeping sickness, WHO goals, elimination of transmission, NTD Modelling Consortium, prediction

This article is included in the 2030 goals for neglected tropical diseases collection.

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Table 1. Summary of modelling perspectives of the WHO goals for gambiense human African trypanosomiasis (gHAT).

<table>
<thead>
<tr>
<th>Current WHO Goal (2020 Goal)</th>
<th>Proposed WHO Goal (2030 Goal)</th>
<th>Is the new target technically feasible under the current disease strategy?</th>
<th>Are current tools able to reliably measure the target?</th>
<th>What are the biggest unknowns?</th>
<th>What are the biggest risks?</th>
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<tr>
<td>Elimination as a public health problem (EPHP). Indicators: (a) &lt;2000 cases globally; and (b) &gt;90% reduction in areas reporting &gt;1 case/10,000 people in 2016–2020 compared to 2000–2004.</td>
<td>Elimination of transmission (EOT). Indicators: (a) zero reported cases; (b) 90% reduction in high and moderate risk areas relative to 2020 baseline; and (c) &gt;50% and &gt;95% of at-risk populations &lt;1 hour and &lt;5 hours from a health facility with gHAT diagnostics, respectively.</td>
<td>The target may be technically feasible using existing tools but perhaps not under the current strategy. EOT may require a step change in the level of surveillance and the use of additional controls (such as door-to-door screening or vector control) in persistent regions. Continued use of existing rapid diagnostic tests, together with 2030 health facility targets, will help case detection. New drugs should improve compliance and ease of treatment.</td>
<td>Existing diagnostics may be sufficient, based on currently reported diagnostic characteristics. However, (i) the indicator of zero reported cases does not imply that the goal of EOT has been reached, (ii) sensitivity could change based on future variation of circulating parasites, and (iii) new tools could improve throughput for large-scale, high-specificity surveillance and/or the ability to detect cryptic human or animal reservoirs.</td>
<td>Prevalence of infection in regions that have never had active surveillance. The role of asymptomatic infections and animal reservoirs as elimination is approached.</td>
<td>Lack of participation in surveillance at a range of scales. Inability to screen and treat due to conflict. Reduction in controls, particularly passive surveillance, once zero cases are reported locally.</td>
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infections for which, through deliberate intervention, global EOT has already been achieved (smallpox and rinderpest) or may be met by 2030 (e.g. Guinea worm and polio).

Predictive, mechanistic modelling is a data-driven approach to explore the feasibility of reaching the WHO goals, taking into account the known biology of infection but also representing uncertainty in all processes. Recent mathematical modelling by the NTD Modelling Consortium and collaborators - including groups from the Institute for Disease Modeling, the Swiss Tropical and Public Health Institute, University of Warwick and Yale University - has provided quantitative perspectives on the challenges of reaching and the likelihood of achieving both the 2020 and 2030 WHO goals for gHAT. The models used have been largely deterministic, which typically comprise of systems of ordinary differential equations (ODEs) and describe average expected infection dynamics, however there has recently been implementation of stochastic models, using Gillespie-based simulation algorithms to simulate the impact of chance events as we approach EOT. Stochastic model results will be identified as such, so other cited modelling studies will use deterministic frameworks. The following sections outline some of the key model findings that are of direct relevance to the 2030 EOT goal.

**Modelling insights from strategies previously conducted**

In the last two decades, the predominant strategy against gHAT was medical only, comprising active screening and passive surveillance followed by treatment. Current medical-based gHAT control strategies are working well in reducing incidence\(^2\) and modelling indicates they are also reducing underlying transmission\(^3\). Shortening time to detection and treatment of cases further reduces morbidity and subsequent onward transmission\(^4\). Modelling indicates that, in Uganda and South Sudan, passive surveillance reduced transmission by 30-50% during the 1990s and 2000s; strengthening these systems in gHAT endemic regions could therefore have great potential\(^5\). Staged gHAT case data (differentiating between stage 1 and stage 2 cases) can provide substantial information on the effectiveness of, and changes in, the passive surveillance system. Usually the proportion of stage 1 cases is low in passive surveillance (~30% in 2012\(^6\)); due to the lack of symptom severity and specificity in stage 1, and thereby limiting the self-presentation of those infected and passive diagnoses made for people in this stage. Conversely, most active detections (mass screening) are in stage 1 (~70% in 2012\(^7\)) as case confirmation relies on serology and parasitology, rather than symptoms. Improvement in time to detection in former Bandundu province in the Democratic Republic of Congo (DRC) is reflected in a greater proportion of stage 1 cases, with modelling estimating a doubling of the stage 1 passive detection rate between 2000-2012\(^8\).

Despite these successes, controls can be disrupted by conflict or other events; notably, the Ebola epidemic in West Africa resulted in temporary cessation of medial activities\(^9\). Furthermore, in higher endemicity settings or regions with little screening, the current medical-only interventions are predicted to be insufficient for achieving EOT by 2030 (e.g. in several health zones in Bandundu, DRC, EOT is predicted to be realised after 2050\(^10\)-\(^14\)). In these settings - assuming scale up of vector control (VC) is feasible and the substantial (>80%) reduction in tsetse density\(^15\)-\(^16\) can be reproduced widely - supplementing medical interventions with VC is predicted to be cost-effective at relatively low willingness-to-pay (WTP) thresholds in high-risk areas\(^14\), and to lead to EOT in much shorter timescales (1–6 years instead of >30 years in some settings)\(^12\),\(^13\). It is noted that deterministic modelling studies are unable to exactly predict when transmission will be eliminated and therefore models have employed a proxy threshold of <1 new infection per 100,000 or 1,000,000 per year. Whilst this proxy is imperfect, more recent stochastic modelling indicates that stochastic and deterministic model dynamics for gHAT follow very similar trends even at low prevalence\(^17\). Furthermore, whilst deterministic modelling may also be unsuitable for some small-scale modelling, stochastic modelling of gHAT in villages finds a population size of around 2,000 is sufficient for persistence, whereas this “critical community size” for persistence of other infections is typically much higher, e.g. around 300,000 people for measles\(^18\); this indicates that deterministic gHAT models at the health zone level (100,000 people) pose a limited cause for concern.

**What are the practical implications of the elimination of transmission goal?**

The WHO 2030 goal for gHAT is EOT globally, with the key proposed indicator of achieving zero reported cases (Table 1). Other proposed indicators relate to sustaining coverage of passive surveillance.

**Measuring the target**

In the long term, reaching EOT will lead to zero detected cases; however, the two objectives are not equivalent - it is possible either that zero case detections could occur without EOT or, conversely, that gHAT detections could be observed even after EOT.

**EOT before zero reporting.** Achieving EOT may not immediately lead to zero detected cases as there is often a long period between infection and detection (several years is typical\(^19\), although in extreme cases this could be decades\(^20\)). As EOT is approached, the choice of confirmatory diagnostics becomes increasingly important as imperfect test specificity, even current, multidagnostic algorithms with ~99.9% specificity (serological testing followed by microscopy), can lead to false positive cases. As we approach the 2030 goal, more rigorous methodologies (e.g. the laboratory-based trypanolysis test with 100% specificity\(^21\)) should help to circumnavigate this problem of positive predictive value.

**Zero reporting but not EOT.** Achieving zero detected cases does not mean that there is EOT for numerous reasons. The first possibility is that screening does not identify all remaining infections at peri-elimination. Only some of the population
at risk is regularly screened; modelling\(^5,6,10\) suggests that some high-risk individuals (~20% of the population) may not attend active screenings and data show that not all settlements in high-risk areas are screened annually (around 50% of villages in a high-endemicity region of DRC were screened in any given year\(^2,16\)). Coverage may improve with mini mobile teams (screening otherwise inaccessible villages) or door-to-door screening (likely increasing the number of high-risk people participating), but pockets of infection could still be missed. Furthermore, large areas of DRC (Figure 1), South Sudan and Central African Republic with potential transmission are not regularly screened due to regional conflicts. Secondly, even where there is a functional health system, there is a high probability of underreporting; models for Bandundu, DRC, suggest that only around 20% of gHAT cases that escape active detection are identified by passive detection (Model W in Castañó et al.\(^{10}\)), corresponding to ~63% of all infections being unreported. Choice and used of available diagnostics are crucial for information certainty – as we approach the endgame it may be that current diagnostics become less able to detect circulating antibodies (due to changing parasite antigen expression) and therefore decreased sensitivity of surveillance tools.

Stochastic models have been used in order to explore the predictive power of one or more years of zero detected cases in estimating the likelihood of EOT at the health zone level (~100,000 people) for the DRC, finding that three years of zero reported cases is sufficient to have >80% positive predictive value (PPV) that EOT has been met\(^1\). Another study with stochastic village-level model (simulating observed active screening from 2000-16 for 559 settlements in Yasa Bonga & Mosango, Bandundu, DRC) also examined detecting zero cases under active screening. This smaller-scale model strongly indicates that three or more consecutive rounds of finding zero cases is sufficient to reach >90% PPV of local EOT across typical village sizes and where screenings that achieve <10% coverage were ignored (Figure 2).\(^4\) There is higher certainty of EOT in smaller settlements and only using active screenings with >50% coverage as a measure could reduce the number of screening rounds needed to have high confidence. Current WHO guidelines recommend conducting three consecutive years of active screening with zero detections in a village before stopping, therefore providing high confidence of local EOT prior to cessation. Modelling suggests that factoring screening coverage and population size into future guidelines could further improve certainty that EOT is met before stopping and could reduce the number of zero detections required for smaller settlements if coverage is sufficient. Scaling these insights between different spatial scales is confounded by spatial correlations and reinvasion, suggesting an intelligent and reactive surveillance methodology is required.

Finally, there is potential for circulation of infection in animal reservoirs or persistence in asymptomatic individuals, which could lead to resurgence even after zero human reporting\(^2,16\) and is discussed in more detail in the “Risks and unknowns” section below.

**Technical feasibility**

Models predict that for some regions (e.g. Equateur, DRC) continuation of the current medical-only strategy could achieve local EOT by 2030\(^2,16\); however, in other regions (particularly some of Bandundu, DRC) this strategy may need to be supplemented with additional or improved interventions, even in areas likely to meet EHPB by 2020\(^1\). Some drivers of local epidemiology that explain these different outcomes include variation in tsetse density and heterogeneity of risk of exposure within regions, in addition to the diversity in implementation and coverage of active screening. Local EOT may be unsustainable without continued control due to the risk of reinvasion from other infected areas.

Multiple modelling approaches have shown that improvements to passive surveillance, targeting active screening to include high-risk groups, and VC could all result in reduced transmission and lead to EOT by 2030 with higher probability than the current medical-only strategy\(^1,16,23\). Whilst it may not be necessary to implement VC across all settings, modelling consistently finds that VC averts infections fastest amongst the considered strategies. Modelling also suggests that the use of...
other new technologies (i.e. new oral drugs and RDTs) could lead to EOT but with lower probability and slower timelines than VC\textsuperscript{14}.

Operational feasibility
Modelling results are generally based on assumptions that the health system retains similar or better functionality over the next 10 years. However, political instability, conflict, or a reduced priority for tackling the reduced number of future cases could all lead to less control being applied in the future. Modelling can also simulate the possible impact of such future disruption to activities (planned or otherwise) in addition to more optimistic assumptions about intervention coverage. Whilst planned cessation following zero reporting has already been considered in some modelling studies\textsuperscript{13,14}, unplanned intervention suspensions and its impact can and should be explored in future modelling work.

Ability to sustain achievement of the goal
Stopping large-scale control activities against gHAT too soon could be problematic for EOT. Modelling was used to explore potential resurgence following attainment of EPHP in Guinea\textsuperscript{15}, concluding that the presence of animal reservoirs would likely lead to resurgence following cessation of screening and vector control, but resurgence was unlikely if transmission was anthroponotic (see “Animal reservoirs” below). Indeed, interruption of medical interventions in Guinea during the Ebola outbreak has shown that early cessation of activities in low prevalence settings can still lead to resurgence over three years\textsuperscript{15}. In contrast, regions that maintained VC but stopped medical intervention observed a decrease in prevalence during the same time period\textsuperscript{15}.

Even if an area has reached local EOT, there is a concern that cessation of activities could be risky if nearby places have ongoing transmission. Stochastic modelling of reinvasion of gHAT in villages in DRC that have achieved local EOT suggests short-term reinvasion is likely (>70\%) from a single infected person, but less likely to cause persistent infection for ~15 years (<20\%). This is due to the high probability that someone will be passively detected and treated or die before creating secondary human infections (through tsetse) even in the absence of active screening. This is also reflected in basic reproduction numbers which only slightly exceed one.

Considerations of costs and allocative efficiency
The burden of NTDs falls in resource-poor settings, and it is of utmost importance to efficiently use the resources available. Combining cost models with dynamic transmission models provides a valuable framework in which to examine the financial and economic impacts and the cost-effectiveness of strategies which account for changing burden as elimination is approached and/or achieved. One such cost-effectiveness analysis for gHAT across settings of different transmission intensities has found that VC combined with other new technologies (diagnostics and drugs) is likely to be highly cost-effective in high-transmission settings (i.e. cost-effective for WTP thresholds >$386/disability-adjusted life year [DALY] averted). In moderate-transmission settings this strategy is only likely to be cost-effective for high WTP thresholds (>$1509/DALY averted), with medical-only strategies using new technologies likely to be preferable for lower WTP thresholds\textsuperscript{14}. Unfortunately, cost-effectiveness in this traditional net-benefits framework does not always align with the goal of EOT by 2030, as strategies that are cost-effective (in terms of
DALYs versus costs) may not be sufficient to meet the EOT goal. For example, in Sutherland et al., VC strategies were generally required to have high predicted probability of EOT by 2030, despite having low probabilities of being cost-effective in moderate- or low-transmission settings.

An analysis on the affordability of gHAT intervention and patient financial impact estimated that the total costs of a global control or elimination programme would be substantial (depending on the programme, between US$410.9 million and US$1.2 billion, compared to US$630.6 million for control activities in 2013-2020). Alleviation of impoverishment and catastrophic health expenditures for households due to gHAT infection can only be achieved through elimination, rather than control, programmes. Ongoing work by the co-authors as part of the HAT Modelling and Economic Predictions for Policy (HAT MEPP) project seeks to assess the cost-effectiveness of elimination strategies based on recent, local data and model updates in order to provide specific and up-to-date recommendations across different settings. It is anticipated that, as before, recommended strategies will not be the same in different transmission settings or geographic regions and will depend on affordability and willingness to pay for averted DALYs or EOT.

**Risks and unknowns faced by gHAT elimination programmes**

A more complete review of key factors that may impact the EOT goal is given by Büscher et al. Here, insights arising from modelling-based studies are discussed.

**Systematic non-participation in screening**

Several modelling studies suggest that there is systematic non-participation in active screening, with high-risk individuals less likely to participate; the models without this heterogeneity were unable to match the observed longitudinal patterns of cases across different regions. More detailed data on age and gender of screening participants and gHAT cases could help better elucidate key groups in the population most responsible for transmission.

**Animal reservoirs**

Although prevalence of gHAT infection in animals is nonzero, estimates are uncertain and the role of infected animals in onward transmission is unclear. One modelling study matched to point prevalence data from animals in Cameroon suggested that animals constitute a possible transmission reservoir, implying that control targeting only human cases would be unable to eliminate gHAT due to persistence in animals, whereas modelling using longitudinal human data (Guinea, DRC, Chad) suggest that there is comparable statistical support for models with and without an animal reservoir. However, animals are unlikely to be able to sustain transmission on their own in Chad. The model used for Cameroon used the next generation matrix approach and the assumption of constant endemic levels of infection, in conjunction with sampled animal prevalences. In contrast the other models only utilised human case data, and fitted to decreasing reporting trends and time-varying intervention activities. Table 2 highlights some future work and data which would be needed to provide greater certainty on animal reservoirs, although it is noted that different geographies may have different potential based on human-tsetse-animal abundance and contact patterns.

The observed decreasing human case trends combined with model fitting to such data provide optimism that there is limited (if any) transmission from non-human animals to humans (through tsetse); however the discovery of transmission cycles in dogs in the last phase of the Guinea worm eradication programme serves as an important reminder that the role of animals should not yet be completely discounted as we aim towards EOT for gHAT. Modelling suggests that VC, including spraying livestock, would reduce any possible transmission from animals, although pockets of sustained transmission could occur away from human activities.

**Asymptomatic reservoirs**

Asymptomatic infections in humans have been considered in a few transmission models. In some, the role of these infections in maintaining transmission or causing resurgence was not directly assessed, although one modelling study using data from Guinea found both asymptomatic and clinical human infections were necessary for gHAT to persist (assuming no animal reservoir) and concluded that passive surveillance alone was not sufficient for gHAT monitoring in the approach to elimination. Generally there is little routinely collected data to help inform asymptomatic transmission modelling. Some case data will contain asymptomatics – while all parasitological-confirmed infections are reported, symptoms are not recorded in aggregated data – but screening diagnostics may be less sensitive on such infections.

**Movement**

So far, little attention has been paid to the movement of people in the modelling literature on gHAT; however, this may be important in areas that recently achieved disease-free status. Particular regions of concern would include formally endemic areas with both high influxes of refugees/interally displaced people and limited surveillance.

**Immediate priorities**

Table 2 highlights a list of priority questions for modellers that are of relevance for the 2030 EOT goal for gHAT arising from discussions between modellers and WHO. In the table the questions are identified from a policy perspective by WHO, and the modellers provide the model and data which would be required to address the policy questions using modelling. Similar tables are provided for other NTDs as part of this special collection.
### Table 2. Immediate priorities for modelling for gambiense human African trypanosomiasis (gHAT).

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<th>Priority issue / question identified by WHO during this meeting</th>
<th>How can modelling address this?</th>
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<td><strong>Probability of interrupted transmission:</strong> Can existing mathematical models be used to define the probability of interruption of gHAT transmission in regions where no cases have been detected?</td>
<td>Using historic data, and assumptions on current passive surveillance, models can be generated that capture the observed dynamics at regional foci and calculate the probability (positive predictive value, PPV) of interrupted transmission given that no cases have been reported for different periods of time.</td>
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<td><strong>Reactive screening:</strong> How does a reactive screening strategy compare to active screening and passive detection, or passive detection alone in terms of: reduction of transmission and associated timescales? case reporting?</td>
<td>Modelers can develop/refine modelling of current active and passive strategies to simulate a reactive screening strategy. - The spatial scale considered will impact results. - Reactive strategies can and should be included in cost predictions and cost-effectiveness analyses.</td>
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<tr>
<td><strong>Animal reservoir:</strong> What do we know about their role in transmitting disease? How could an animal reservoir affect the 2030 target?</td>
<td>Some modelling has already explored possible animal reservoirs. Modelers can continue to explore: - Whether there are signals of animal reservoirs by assessing human case data alone - If there is any support for these models, to assess the relative contribution of animals to transmission, and what impact this could have on timescales to achieve EOT - To include animals in a village-scale model framework (to assess PPV of zero case detections in active screening on EOT) - To make estimates more robust by fitting to human and animal data (if available) - To assess implications of animal reservoirs in decision analyses between interventions.</td>
</tr>
<tr>
<td><strong>Asymptomatics:</strong> Can we estimate the potential number of asymptomatic infections? E.g. for one detected case, how many remain undetected? How likely are asymptomatics to infect others? What do we know about their role in (maintaining) transmission?</td>
<td>- Existing modelling frameworks can be adapted to include potential asymptomatics (including self-cure or skin infections) - Sensitivity analysis and/or matching to data (if available) could estimate possible numbers of asymptomatics, their relative contribution to transmission, infection timescales, and relative infectivity. Lack of data may lead to large confidence intervals - Modellers can evaluate the effectiveness of different strategy types in models with and without asymptomatic people - e.g. would we select the same intervention strategy if asymptomatics play a substantial role in transmission?</td>
</tr>
<tr>
<td><strong>Spatial prediction:</strong> Support defining areas that should be screened, where there is potential of transmission. Similarly, can we rule out certain areas?</td>
<td>- A tsetse absence model could be used to assess regions which are unlikely to have gHAT due to unsuitable habitat. - This can be used to explore the joint distribution of the active and passive surveillance data and to look for factors/variables which could predict the underlying variation and probability of reporting. - It may be possible to include a range of factors into these predictions including changing population distribution and land-use.</td>
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### Data availability
No data are associated with this article.

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References

1. Global Health Observatory data repository [Internet]. Reference Source
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I have read through the revised article and think the authors have done a great job responding to my original comments and improving the manuscript accordingly. I’m happy with the current version to be accepted for indexing.

Competing Interests: No competing interests were disclosed.

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

Version 1

Reviewer Report 07 November 2019

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This open letter provides an overview of the use of mathematical modelling of gHAT and its uses in support of the WHO's target of global elimination of transmission by 2030. Coming from a modelling background but with little knowledge of gHAT I read this letter with much interest. However, what I feel is currently lacking is a bit more detail and structure regarding the models
and their predictions themselves. That is, from the article it is clear that various models have been
developed and that they have been applied to answer slightly different gHAT-related questions.
What is not clear is what types of models have been used for what questions, where do models
generally agree or disagree, what are crucial knowledge gaps highlighted by the these exercises,
and what is the direction that models should be heading to support local and global elimination
efforts. I believe that much of this could be achieved simply by re-organisation/restructuring,
making the sections more focused and less overlapping. For example, the problems of a potential
animal reservoirs or asymptomatic infections come up more than once, the same with
screening/diagnostics. Personally I would focus on just a few key obstacles, important aspects of
gHAT epidemiology/elimination, or intervention measures, detailing the problem and then
illustrating what modelling has found and/or how it can be used in the future. Also, the authors
mention in the Abstract that accounting for uncertainties and stochastic effects is very important,
which I fully agree with, especially as one approaches disease elimination scenarios. However, I
was missing the discussion on this in the text; are none of the approaches developed so far
stochastic/deal with uncertainty? Given its importance I would suggest that the authors could
maybe dedicating a separate (sub-)section on this?

More specific comments:
- Background, second paragraph: maybe this is a bit negative but saying that 963 cases
  suggests that the first indication has been greatly surpassed the goal of getting it below
  2000 is a bit of an exaggeration; to me this suggests things are on target
- Modelling insights from […], first paragraph: it is not immediately obvious how staged gHAT
  case data can provide substantial information on the effectiveness of surveillance; more
detail on this would be welcome
- Table 1: the question ‘is the new target technical feasible […]’ is answered by ‘yes’; but the
  next row goes into ‘if not, what is required’ - given that the first answer is yes, do we need
  this then? Or is the answer not really ‘yes’
- EOT before zero reporting: what are ‘algorithms’ with 99.9% specificity?
- Zero reporting but not EOT: this seems to me like a two-part problem, the first is general
  under-reporting, and the other long periods of asymptomatic infections; would it be worth
  making this distinction?
- Zero reporting but not EOT, second paragraph: replace ‘replaying’ with ‘simulating’
- Technical feasibility: explain why the strategy could achieve local EOT in some regions but
  not in others
- Ability to sustain achievement […]: it seems like the issue about the potential animal
  reservoir is slightly controversial, i.e. what's the evidence for or against?
- Ability to sustain achievement […], last paragraph: why is reintroduction less likely to cause
  persistent infections? Presumably control programs would stop after elimination has been
  achieved?
Asymptomatic reservoirs: this is a clear example where it would be good to have more of an overall discussion on where models agree, where they differ and why.

Table 2: do you need the second column, or would it be possible to have this table just listing/detailing (top priority) questions that can be addressed/answered with modelling? Personally I find that the 'how' is not that important, especially as no other model details and assumptions are being discussed here

Overall I think that this is an important piece of work highlighting the uses of mathematical models in public health, and a little more structured and focused approach would make it even better.

Is the rationale for the Open Letter provided in sufficient detail?
Yes

Does the article adequately reference differing views and opinions?
Yes

Are all factual statements correct, and are statements and arguments made adequately supported by citations?
Yes

Is the Open Letter written in accessible language?
Yes

Where applicable, are recommendations and next steps explained clearly for others to follow?
Yes

**Competing Interests:** No competing interests were disclosed.

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard, however I have significant reservations, as outlined above.

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**Author Response 02 Apr 2020**

**Kat Rock,**

Thank you for your helpful comments, we have added several clarifications to the manuscript as suggested (see below), and although we do not consider that complete article restructuring is necessary we have included more cross-referencing to link different sections of the article.

- To improve our clarity about model types we have firstly added the following text to the background section: “The models used have been largely deterministic, which typically comprise of systems of ordinary differential equations (ODEs) and describe average expected infection dynamics, however there has recently been implementation of stochastic models, using Gillespie-based simulation algorithms to
simulate the impact of chance events as we approach EOT.”

We also note the proxy thresholds required if assessing predictions of EOT in deterministic frameworks in our “Modelling insights” section: “It is noted that deterministic modelling studies are unable to exactly predict when transmission will be eliminated and therefore models have employed a proxy threshold of <1 new infection per 100,000 or 1,000,000 per year. Whilst this proxy is imperfect, more recent stochastic modelling indicates that stochastic and deterministic model dynamics for gHAT follow very similar trends even at low prevalence (18). Furthermore, whilst deterministic modelling may also be unsuitable for some small-scale modelling, stochastic modelling of gHAT in villages finds a population size of around 2000 is sufficient for persistence, whereas this “critical community size” for persistence of other infections is typically much higher e.g. around 300,000 people for measles (19); this indicates that deterministic gHAT models at the health zone level (100,000 people) pose limited cause for concern.”

To clarify the importance of staged case data we have added: “Usually the proportion of stage 1 cases is low in passive surveillance (~30% in 2012 (10)); due to the lack of symptom severity and specificity in stage 1, and thereby limiting the self-presentation of those infected and passive diagnoses made for people in this stage. Conversely, most active detections (mass screening) are in stage 1 (~70% in 2012 (11)) as case confirmation relies on serology and parasitology, rather than symptoms. Improvement in time to detection in former Bandundu province in the Democratic Republic of the Congo (DRC) is reflected in a greater proportion of stage 1 cases, with modelling estimating a possible doubling of the stage 1 passive detection rate between 2000-2012 (9).”

○ In Table 1 we have moved some of the technical feasibility text into the first box, although our uncertainty about strategies rather than tools themselves means we have left creation of novel screening strategies in the “if not” box.

○ The simulations for reintroduction include cessation of screening activities, however very low R_0 values (just above 1) mean that local extinction is likely from a single case. We added: “This is due to the high probability that someone will be passively detected and treated or die before creating secondary human infections (through tsetse) even in the absence of active screening; this is also reflected in basic reproduction numbers which only slightly exceed one.”

○ For asymptomatic reservoirs we change “unclear” to “not directly assessed” to highlight that the models don't necessarily disagree, but this has not generally been explored. We also add an additional sentence to explain why the issue of asymptomatics has been studied infrequently.

○ We would like to retain the second column in Table 2 for two reasons. Firstly, the questions themselves were highlighted by WHO as priorities, irrespective of whether modelling and data are at a suitable stage to answer them. Our perspective on the data needs and modelling requirements to answer them indicate how readily modelling can be used to address these questions. Secondly, this letter is part of a special collection in which a similar table is provided for each NTD. We have added additional text in the immediate priorities section to explain the table more clearly.

**Competing Interests:** No competing interests were disclosed.
The open letter is a clear overview of modelling efforts tracing the road to elimination of gambiense HAT.

- Background: Paragraph 1, line 6: note parasites can cross the BBB before establishing themselves there to make stage 2 infection, so please change to reflect that.
- Line 11: After 37,385 - add the word "reported" before cases.
- Paragraph 2, line 4: 2,000 (add the comma). Line 9: Also add a comma to 1,419.
- Paragraph 3, line 10-11: place commas around "and likelihood of reaching".
- Table 1, point 5: I am not certain that "existing diagnostics are likely sufficient". If skin parasites are not sustaining serum antibody titres, and also if the standard antigens used in current tests are not present in a cohort of residual parasites it is likely the diagnostics will become ever less sensitive.
- On page 4: the reference to the trypanolysis test which purports 100% specificity is, I think, equally vulnerable to loss of sensitivity in cases where parasites not expressing antigens central to this test are in circulation.
- On page 5: Second paragraph points to the potential of animal reservoirs and asymptomatic individuals. However, the downward trend appears to have been following predictions quite nicely (where these refugia are not considered). Does the modelling so far, therefore, rule out a significant impact of the animal reservoir and asymptomatic patients, or does incidence have to get even lower before the problem will become manifest?
- Page 5. section Operational feasibility: It is stated that "Models are therefore making assumptions that screening and other controls follow recent trends." Can't the models themselves be used now to predict what happens in the event of different removal of control scenarios?
- Page 6, Last paragraph before the "Risks and unknowns....." section: This covers estimate of cost of control based on the Sutherland and Tediosi work. Clearly these estimates could have a large impact on policy decisions. Can the authors offer an opinion on how robust they consider those findings (in the light of modelling predictions?)
Page 6, column 2, section on "Movement", line 4: "formally" should be "formerly"

References, reference 1: More information needed to access.

Some references e.g. 4, 12, 14, 15, 20, 21, 26 are using capital first letters for individual words in the title, while others are not.

**Is the rationale for the Open Letter provided in sufficient detail?**
Yes

**Does the article adequately reference differing views and opinions?**
Yes

**Are all factual statements correct, and are statements and arguments made adequately supported by citations?**
Yes

**Is the Open Letter written in accessible language?**
Yes

**Where applicable, are recommendations and next steps explained clearly for others to follow?**
Yes

**Competing Interests:** No competing interests were disclosed.

**Reviewer Expertise:** African trypanosomiasis, drugs mode of action and resistance mechanisms

I confirm that I have read this submission and believe that I have an appropriate level of expertise to confirm that it is of an acceptable scientific standard.

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**Author Response 02 Apr 2020**

**Kat Rock,**

Thank you for your helpful comments, we have added additional clarifications to the manuscript as recommended:

- We agree we should be more careful with our wording in Table 1 about diagnostics. We have updated the text to: “Existing diagnostics may be sufficient, based on currently reported diagnostic characteristics. However, (i) the indicator of zero reported cases does not imply that the goal of EOT has been reached, (ii) sensitivity could change based on future variation of circulating parasites, and (iii) new tools could improve throughput for large-scale, high-specificity surveillance and/or the ability to detect cryptic human or animal reservoirs.”
- To address the possibility for decreasing sensitivity of diagnostics we added in the “Zero reporting but not EOT” section: “Choice and use of available diagnostics are
crucial for information certainly – as we approach the endgame it may be that current diagnostics become less able to detect circulating antibodies (due to changing parasite antigen expression) and therefore decrease sensitivity of surveillance tools.”

- We discuss animal reservoirs primarily on page 6 (whilst mentioning them on page 5) so have now added a cross-reference “and is discussed in more detail in the “Risks and Unknowns” section below.” In that section we add a note on decreasing case trends increasing our optimism: “The observed decreasing human case trends combined with model fitting to such data provide optimism that there is limited, if any, transmission from non-human animals to humans (via tsetse), however the discovery of transmission cycles in dogs in the last phase of Guinea worm eradication programme (26) serves as an important reminder that the role of animals should not yet be completely discounted as we aim towards EOT for gHAT.”

- You are correct that models can be used to predict cessation or removal of controls and we have now added the following: “Modelling can also simulate the possible impact of such future disruption to activities (planned or otherwise) in addition to more optimistic assumptions about intervention coverage. Whilst planned cessation following zero reporting has been already considered in some modelling studies (Davis, 2019), unplanned intervention suspension and its impact can and should be explored in future modelling work “

- Firstly, we would like to emphasise that the cited cost-effectiveness study did utilise both a cost and transmission dynamic modeling framework, therefore it accounted for a decreasing burden as predicted by the transmission model for different interventions. (We have added a sentence to clarify this - “Combining cost models with dynamic transmission models provides a valuable framework in which to examine the financial and economic impacts and the cost-effectiveness of strategies which account for changing burden as elimination is approached and/or achieved. One such cost-effectiveness analysis...”) We also added some text on our ongoing work which highlights important analyses for providing updated and localised strategy recommendations although we believe the overall message will hold: “Ongoing work by the co-authors as part of the HAT Modelling and Economic Predictions for Policy (HAT MEPP) project seeks to assess the cost-effectiveness of elimination strategies based on recent, local data and model updates in order to provide specific and up-to-date recommendations across different settings. It is anticipated that, as before, recommended strategies will not be the same in different transmission settings or geographic regions and will depend on affordability and willingness to pay for averted DALYS or EOT.”

**Competing Interests:** No competing interests were disclosed.