Monitoring and evaluation of malaria-related routine data during the COVID-19 pandemic

August 2020
**Background**

Routine disease surveillance will be critical for monitoring where malaria control, elimination and prevention of re-establishment efforts may be faltering during the pandemic as well as where COVID-19 may be spreading. Monitoring and evaluating key routine health indicators will help address key questions including:

- How are the systems that diagnose, treat, and report malaria being affected by COVID-19?
- Are unusual changes in malaria incidence occurring? If so, are these changes influenced by COVID-19?
- Are non-malaria fevers, which may be attributable to COVID-19, increasing?
- Are lessons being learnt that may inform short, medium, and long-term actions for malaria surveillance?

Most countries already have such systems in place and are regularly monitoring the situation. Leveraging existing systems to monitoring a set of disease indicators throughout the pandemic will help countries better understand whether malaria case management services continue to operate, where disruptions to service provision or care seeking may have occurred, where unexplained fevers could potentially be due to increased incidence of COVID-19 or other febrile diseases, where both COVID-19 and malaria may be circulating, and what actions may be put in place as a mitigation plan.

**Approach**

We will describe four main steps to answer these questions as illustrated below.

- **Step 1:** Continuing to monitor key malaria (or malaria related) indicators to assess where disruptions to access to care and to health service provision may be occurring.
- **Step 2:** Analyzing and interpreting the indicators.
- **Step 3:** Identification of root causes and other contributing factors for the interpretation of key malaria indicators.
- **Step 4:** Tailoring the actions to the root causes and observed malaria trends.
Figure 1 above shows the flow of the steps from monitoring to tailoring interventions. In every step data quality should be assessed and checks should be performed to ensure that you are seeing true trends and not just missing or incomplete data.

**Step 1: Continuing to monitor key malaria (or malaria related) indicators**

This document provides guidance on how to assess where disruptions to access to care and to health service provision may be occurring, where malaria control, elimination or prevention of re-establishment may be faltering, or where COVID-19 may be spreading.

Most indicators that you would use to assess these trends (Table 1) should already be monitored through routine health systems, although some may not always be collected and/or available in every country. The sources (and quality) of data for the indicators may vary but usually exist through a monthly reporting system (e.g. Health Management Information System - HMIS), weekly reporting system (e.g. Integrated Disease Surveillance and Response - IDSR), or program activity reports. Some new data reporting systems may have been established for COVID-19 specifically and may provide valuable information to help interpret the malaria metrics (e.g. febrile cases may not be currently collected in HMIS or IDSR but may be in new COVID-19 specific systems, or they may be collected but not properly reported yet).

Most of these indicators should already be monitored and can continue to be monitored in the absence of major COVID-19 disruptions. Under the scenario of major disruptions in health services within a COVID-19 emergency response, however, it may not be feasible to collect some of the
indicators. For example, an indicator may not be available (due to a lack of reporting) or may be less relevant (if programs shift to treatment of presumptive cases). Despite these challenges, maintaining the indicators as part of ongoing data collection efforts throughout this period is important. When health services and the reporting system(s) are operational again, key indicators to monitor will include health facility reporting and form completeness to provide insights into the feasibility of leveraging routinely collected data. Working with partners and documenting when and where these disruptions start will help in data interpretation in the short and long term.

Table 1: Key routine malaria-related indicators

<table>
<thead>
<tr>
<th>Dimensions of change</th>
<th>Potential malaria-related indicators¹</th>
<th>Impact and interpretation²</th>
<th>Relevance of the indicator during COVID-19 disruptions</th>
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</thead>
</table>
| **Availability of commodities:**          | Consider the Laboratory Management Information System - LMIS as a data source if available in the country. | Percentage of facilities that had a stockout of commodities in the last reporting period. | If there are stockouts, case management may be impacted (due to low testing) and may result in an increase in malaria cases (due to stockout of treatment) and/or increase in reporting of clinical/suspected malaria cases. Root causes for commodity stockout may involve different levels of the supply chain:  
  - Lack of supplies at the manufacturer level.  
  - Procurement of commodities from out of the country being impacted by overseas COVID-19 policies and lockdown procedures. |
| **Availability of essential malaria commodities** | Evaluate whether rapid diagnostic tests (RDTs) and malaria medicines/drugs continue to reach points of care, or whether COVID-19 has affected supply chains. | Percentage of community health workers (CHWs) who had a stockout of commodities in the last reporting period, by type of commodity. | If points of care stop collecting or reporting indicators because of COVID-19 disruptions, then CHW/facility-level stockout information will not be available and other sources of data will be necessary (e.g. procurement data). |

¹ Specific indicators may vary between countries, and they may not be collected routinely in current health or disease information systems. Further guidelines and lists of indicators are available in *Malaria Surveillance, Monitoring & Evaluation: A Reference Manual* (Geneva: World Health Organization; 2018).

² Guidance on exploring temporal trends (i.e. whether the indicators increase or decrease over time) is provided in Step 2 below “Analyzing and interpreting the indicators.”
<table>
<thead>
<tr>
<th>Uptake of health care services: Age-specific and sector (public, private, community) disaggregation is critical in this context.</th>
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<tbody>
<tr>
<td><strong>Health-seeking behavior and access to health care services</strong></td>
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<tr>
<td>Monitor whether all-cause health care utilization is decreasing (or increasing) as a consequence of COVID-19 interventions, disruptions, or epidemiology.</td>
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</table>
COVID-19 messaging encouraging people with fevers to stay at home, etc.

**Reporting:** Information system (source may be HMIS, IDSR, or other).

<table>
<thead>
<tr>
<th>Data quality: reporting rates and timeliness of reporting</th>
<th>Number (percentage) of facilities (or other points of care) that submit reports, and submit within the required deadline.</th>
<th>Low reporting (notification) rates may impact reported malaria metrics (e.g. number of malaria cases, which may falsely appear to be lower) and bias their interpretation, leading to inadequate conclusions on the malaria situation.</th>
</tr>
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<tbody>
<tr>
<td></td>
<td>In elimination settings the number (percentage) of cases that have been notified within required time, as per country protocol</td>
<td>Low reporting rates themselves may be an indication of problems with health administration or closures due to perceived COVID-19 exposure risks and/or lack of personal protective equipment (PPE). To assess recent change, the percent of reports on time may be more useful, as this reflects reporting only for the most-recent month. The importance of this measure is limited by the proportion of facilities that typically report on time.</td>
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<tr>
<td></td>
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<td>If points of care stop collecting or reporting indicators because of COVID-19 disruptions, then other sources of data will be necessary. It will be important to create linkages with the COVID-19 response team to determine if reporting on malaria testing is occurring through another data flow sequence.</td>
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<tr>
<th>Data quality: form completeness</th>
<th>Percentage of forms that are</th>
<th>Low completeness and or</th>
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Monitor the trends in completeness rate of forms from health facilities or other points of care.

Data quality: case investigation and classification

Monitor the trends in completeness of case investigation and case classification

In elimination settings – percentage of cases that are investigated and classified

Investigation rates may impact the quality of the reported malaria metrics (e.g. key information may be missing) and bias their interpretation, leading to adequate conclusions on the malaria situation.

Investigations) are too low in a specific area or from specific health facilities due to COVID-19 disruptions, then interpretation of other metrics will be limited and other sources of data would be necessary.

In elimination settings, it becomes difficult to classify cases without complete case investigation. Interim case investigations could be done through phone call where possible.

Outcome: Case management

Age-specific and urban/rural disaggregation is critical in this context.

Testing

Monitor the quality of case management services.

Number of malaria tests performed.

Testing rate: Number of malaria tests out of number of fever and/or suspected cases of malaria.

If testing rates in normal clinic flow decrease, it may be difficult to see patterns of non-malaria fevers. Low testing rate may be the consequence of stockout of commodities (or of PPE to safely perform the tests) and/or of low health care utilization (e.g. fewer people with fevers are seeking care due to COVID-19 messaging).

An increase in testing may reflect an increase in suspected malaria cases (i.e. febrile cases) linked to an actual increase in malaria cases, COVID-19, or any other

There may be no testing done because of stockouts, lack of PPE for health care workers, or changes in testing guidelines, in which case the denominator of some indicators (e.g. treatment and incidence) will include clinical malaria cases or will not be estimated (e.g. test positivity rate). Key secondary indicators could be used for interpretation (e.g. changes in number of suspected/clinical cases).
<p>| <strong>Treatment</strong> | Number of malaria cases treated out of the number of confirmed malaria cases diagnosed and/or of suspected and “clinical” malaria cases. | Treatment may be impacted by supply chain challenges or changing treatment practices at points of care or low health care utilization. Low treatment rate may eventually impact malaria trends (e.g. increased malaria transmission). High rates of treatment may be observed in the absence of malaria diagnosis due to a potential increase in the presumptive treatment of any suspected malaria/febrile cases. | The denominator may include suspected malaria cases if confirmatory tests are not available and/or used. Monitoring the proportion of malaria cases treated without confirmation could inform any changes in practice due to COVID-19. |
| <strong>Preventive treatment</strong> (intermittent preventive treatment in pregnancy [IPTp]) | Proportion of pregnant women who received three or more doses of IPTp. | Decrease may be due to challenges around commodity stockout or decreased antenatal care (ANC) visits (e.g. clinic curfews under government COVID-19 policies). | As first ANC visit attendees are often used as the denominator for measuring IPTp coverage, tracking trends in ANC attendance is essential. If COVID-19 disrupts attendance, the estimated IPTp coverage will be affected. |
| <strong>Impact: Malaria trends</strong> | | | Age-specific and urban/rural disaggregation is critical in this context, including vulnerable groups if possible. |
| <strong>Malaria incidence</strong> | Number of confirmed outpatient diagnoses of malaria out of estimated total population of areas | An increase may show an increase in malaria transmission and/or the presence of malaria outbreaks | Malaria incidence Measure changes in malaria |</p>
<table>
<thead>
<tr>
<th>Measure</th>
<th>Description</th>
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<tbody>
<tr>
<td>Trends in febrile illnesses</td>
<td>Number of fever or suspected malaria cases. Could be proxy indicators of illness (e.g. increase in febrile cases may be due to malaria,</td>
</tr>
<tr>
<td>Trends in malaria test positivity rate</td>
<td>Number of positive tests out of the total number of malaria tests. May be a useful measure to show if malaria is increasing in a certain region, assuming testing practices have not been affected. An increase in test positivity rate may suggest a failure of malaria control measures. A decrease in test positivity rate may be the result of an increase in non-malaria fevers such as COVID-19. An increasing fever rate in the absence of a decreasing test positivity rate might suggest that fevers occurring due to COVID-19 and malaria may be simultaneously increasing (in which case test positivity rate may remain stable).</td>
</tr>
<tr>
<td>Test positivity rate</td>
<td>Measure changes in malaria trends.</td>
</tr>
<tr>
<td>Trends in febrile illnesses</td>
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As an example. If you explore whether there has been a reduction in treatment seeking by visualising the reduction in number of OPD visits compared to values in previous years. First ensure that any trends you see are not based on incomplete reporting or a delay in reporting by applying checks on delays and reporting rates.

**Step 2: Analyzing and interpreting the indicators**

Indicators should be interpreted together to understand the impact of COVID-19 on malaria care and preventive services and epidemiology. Caution must be exercised when interpreting these indicators; because of the reporting time lag inherent in most countries (several months are often required for facilities to report), the reported estimates for the most recent months may typically look low, and any changes in estimates over the most recent three months may be unreliable.

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Figure 2: Checking OPD trends against reporting completeness and timeliness

Check 1: is this reduction a result of a delay in reporting?

Check 2: If there is a delay in reporting is this delay unusual? Examine number of reports on time to see if there is less reporting than usual.

Further investigation required to confirm if this reduction is due to reporting delays or changes in treatment seeking behaviour due to COVID

OPD visits were significantly lower in June

No change in the number of timely reports in June indicating no underlying change in reporting

If data viewed before the deadline of report submission or for on time reporting this value will be low.
Figure 2 above shows the steps in visualize a time series graph of OPD visits and performing checks to understand if the trend of the graph is a true trend or is due to the number or delay in health facilities that are reporting data.

There are a few factors to consider:

- **Spatial resolution level**: These metrics should be monitored together at a disaggregated level; district or facility level will be most useful for identifying problems in specific geographic regions.
- **Frequency**: These metrics should be monitored together on a monthly or weekly basis depending on the frequency of reporting in the system and compared against historical data from the same time of year (over at least three prior years) to detect any unusual changes in malaria epidemiology or case management patterns.
- **Disaggregation levels**: Indicators should be monitored at different levels including the following disaggregation factors: age, urban/rural, and point of care (e.g. community vs health facility). Further suggestions are provided in Table 1.
- **Point of care stratification**: Further analysis should be performed at the point of care level to identify which ones may be driving the observed trends (e.g. OPD decline, fever case increase, confirmed malaria case increase...) and target the response to the points of care of major importance (e.g. those with high number of outpatient).
- **Quality of the data and limitations**: Indicators should be estimated from high quality data that are complete enough (thresholds to inform whether data are high quality may vary from one country to another). Each indicator should also be interpreted against potential limitations (e.g. some infections may be asymptomatic and will not be reflected in febrile case trends). Trends can be interpreted from data of acceptable quality that is reported consistently.
- **Visualizations and analysis**: should include temporal trends over a recommended period of at least three years (however the time period may be shorter if high quality data are only available for fewer than 3 years), by month, to compare seasonal changes. Ideally, these temporal trends can be automatically generated within accessible and user-friendly dashboards (e.g. within District Health Information System - DHIS2).
  
  **Anomaly detection algorithms**, as detailed in WHO's *Malaria Surveillance, Monitoring & Evaluation* reference manual, can help detect any significant increase and/or decrease in the monitored indicators. There are other analytical methods such as interrupted times-series analyses to detect any significant changes over time. The observed trends may trigger further analysis to identify the factor(s) behind an increase or decrease of the indicator compared to previous years according to the disaggregation levels suggested above (e.g. low testing rate would require exploring availability and use of RDTs or whether there has been a policy shift to presumptive treatment).

The line graph in Figure 2 can be easily generated for many indicators. However, once an investigation is launched, disaggregating the time-trend analysis by geography in a stepwise manner from the national level to regional, district and facility levels is an important step to determine WHERE the disruption/change has occurred. At each step, ask WHERE did the disruption/change take place? This disaggregation can take place at:

- Individual **health facilities** (as shown in the heat map of Figure 3)
- **Urban versus rural** facilities
• Public versus private facilities
• Primary, secondary, tertiary, or community level

Identifying WHERE the disruption/change occurred will help you understand whether the issue is clustered in particular areas and to tailor further investigation to identify root causes (local and systemic) and to identify appropriate corrective actions. Generating heat maps, as show in Figure 2, can give you a quick sense of whether there a spatial pattern to the disruption/change, so that you know where to focus further investigation or intervention. If indicators for change over time (e.g. change in OPD) are not available, you can compare two or map to determine WHERE the disruption/change has occurred over time – and to identify spatial patterns in the disruption/change over time.

Figure 3: Heat map showing the spatial patterning of changes in OPD as reported by health facilities

Figures 2 and 3 show examples of visualizations of temporal trends and spatial patterns in malaria indicators: total outpatient visits (OPD). This has been generated both as line a graph compared with other years as well as a heat map showing the gradations of the changes. The observed temporal trends within and across geographies should be overlaid with the timelines of COVID-19 interventions (e.g. lockdown) and/or potential disruption (e.g. RDT stockout) for interpretation.
If you have software that allows to easily choose between sub-district designations (such as health facility) as is shown in Figure 3, you could look at multiple indicators for a sub-district area.

**Figure 4: Line graphs generated by Tableau showing trends in multiple indicators over time**

Step 3: Identification of root causes and other contributing factors for the interpretation of key malaria indicators

Root cause analysis is a method for identifying the “root” or main cause(s) of a problem. It typically involves three steps:

- Problem identification: The problem should be identified and clearly stated, for example: “OPD attendance was lower in April 2020 compared to the same period in previous years”.
● Identification of possible causes: Possible causes are factors that may contribute to the problem but are not root causes.

● Differentiation of root causes from possible causes: Root causes are the main causes of the problem. If removed, they eliminate the problem. Note: problems can have more than one root cause. Root cause analysis often uses tools such as root cause trees, 5 Whys, fishbone diagrams and/or patient fall-out analyses to differentiate root causes from possible causes.3

In the present context, there are several areas that could impact malaria services and would need to be explored, including:

● **COVID-19 interventions**: Are/were lock-down measures and travel restrictions put into place, since when/for how long, are/were these deployed in specific areas only or nationwide? How well are/were they respected? Measures may impact the malaria situation negatively (e.g. people would be less likely to seek care) or positively (e.g. people staying at home because of curfew or lock down may be less exposed to mosquito bites in geographies where vectors bite mainly outside and/or where people may be more receptive to media campaigns).

● **Health care utilization**: Are/were policies put into place that could directly impact health care utilization (e.g. access to health care for routine non-emergency situations is prohibited or disincentivized).

● **Malaria surveillance**: Are/were policies put in place that could impact reporting of malaria data (e.g. changes in fever case management such as changes in diagnostic algorithm for malaria), sharing of information (e.g. lack of transparency), investigation of cases (e.g., change to interviews by phone rather than in person) or analyzing the information (e.g. data review meetings cancelled because gatherings are prohibited).

● **Disruption in malaria prevention campaigns** (e.g. seasonal malaria chemoprevention or vector control interventions such as indoor residual spraying or insecticide-treated nets) and/or in **delivery of care in the community** (e.g. testing and treatment provided by community health workers). *Corresponding trackers of COVID-19 impact may already exist and will help interpret the routinely collected metrics.*

● **Other factors**: For example, the spread of COVID-19 impacts economic activities, which may trigger movement of economic migrants and consequently affect access to routine health care services.

Interventions along with changes in policies and guidelines may explain and/or influence trends in service delivery, case management, and reporting practices and will provide the foundation to interpret the key malaria indicators already collected in current health information systems as detailed in Step 1. There may be additional indicators not routinely collected through current systems that would provide valuable context around the interpretation (e.g. availability of human resources to understand if there are enough health workers to provide services at facility or community level, report cases, conduct supervision visits, and monitor disease trends).

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Non-malaria OPD decreases | Non-malaria OPD increase
---|---
Malaria case counts decrease | Are people avoiding all care seeking? | Are there other outbreaks with fever symptoms (possible COVID-19)?
Malaria case counts increase | Is there a malaria outbreak? Are people only seeking care when very sick? | Is there an increase in malaria testing of non-malaria fevers amongst asymptomatic people?

Further contextual information, and other data sources would be important to identify the drivers behind the observed trends. For example, it would be important to these changes alongside stock-out data for these health facilities to ascertain first if changes in TPR are due to lack of available tests rather than underlying causes of fever and health seeking behavior. In addition, it would be important to analyze travel orders or local lockdown measures to interpret changes in OPD attendance. If stockouts have not been an issue, and guidance allows for travel to health facility, further investigation could follow patterns seen in the table above.

Step 4: Tailoring the actions to the root causes and observed malaria trends

The potential actions detailed for each routine indicator will be specific to each country and each context. In general, the root causes driving the observed trends should be explored and addressed. Corrective actions should be monitored (what, by whom, where and by when) and information on what worked / didn't work to resolve the root cause(s) should feedback to refine the corrective actions. Maintaining communication with health care workers at the facilities or within the communities and at different levels of the health systems (e.g. district malaria focal points) will be key to interpreting the observed pattern for each indicator and identifying possible actions. For example, low testing may be due to true stockout or it may reflect that data on stockouts are not reported. Any challenge with stockout of commodities should trigger an investigation into what level of the supply chain is affected (e.g. manufacturer level, international distribution, supply to health facilities) and whether the issue has an effect on all regions in a country or is geographically concentrated. Challenges in terms of data reporting may trigger a simplification of the surveillance workflows (e.g. decreasing numbers of indicators) and/or a supervision visit (potentially virtual) to the underperforming facilities or sensitization of the health staff to reporting. An increase in the number of malaria cases may trigger some reactive measures used to control malaria outbreaks. Actions may differ based on the COVID-19 response phases and also may vary across different geographies within the country. Table 2 provides detailed potential actions depending on observed trends. More information about actions can be found in Tailoring Malaria Interventions in the COVID-19 Response (Geneva: WHO; 2020).
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<thead>
<tr>
<th>Dimensions of change</th>
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<tr>
<td><strong>Availability of commodities</strong></td>
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| Availability of essential malaria commodities (i.e. rapid diagnostic tests [RDTs] and malaria treatment) | If there are changes in availability, explore which level of the supply chain is affected and the root causes, and address them. For example:  
**If there is a shortage of RDTs in the country**  
- Analyze whether the shortage is due to a greater demand and re-evaluate provincial/district restocking and forecasting.  
- Keep records of where shortages of RDTs occur and for how long.  
- The WHO provides guidance (Tailoring Malaria Interventions in the COVID-19 Response) on extraordinary measures that can be implemented under exceptional situations only (e.g. presumptive treatment of fever).  
**If there is a shortage of ACTs in the country**  
- Analyze whether artemisinin-based combination therapy (ACT) consumption is rising due to testing suspected COVID-19 cases for malaria.  
- Determine whether provincial/district restocking and forecasting need to be addressed.  
- Keep records of where presumptive treatment or shortage of ACTs occurs and for how long. |
| **Uptake of health care services** | |
| Health seeking behavior and access to health care services | Any increase in OPD attendance, particularly for fever or flu symptoms, should be communicated to the COVID-19 response team.  
If OPD attendance decreases, there will likely be a decrease in total malaria cases reported. COVID-19 communication campaigns should include messages to continue to use other health services and describe the actions taken to protect patient safety to ensure that those most vulnerable to malaria morbidity and mortality (for example, children under 5) continue to seek care.  
Note any shifts in case loads reported by community health workers. This could indicate a shift from health service utilization at the facility level to the community. |

**Reporting**: Information system (source may be HMIS, IDSR, or other).
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| **Data quality: reporting**                              | A decrease in reporting rate or completeness may suggest deteriorations in the health system for which root causes need to be addressed (e.g. overburdened staff due to increased case reporting, health care workers not at work, and/or additional reporting responsibilities for COVID-19 lead to struggling to collect and report information). Specific actions may include:  
  ● Adaptations to data flow: e.g. concentrate reporting on minimum essential indicators, potentially bypass some administrative hierarchies.  
  ● Substitute/switch to mobile/electronic reporting if appropriate.  
  ● Increase general awareness of the importance of reporting.  
  ● Target PoCs that report less and are historically driving most of the patient case load with specific supervision and messaging around malaria reporting. |
| **Data quality: form completeness**                       | The COVID-19 response team may set up reporting systems that are different in their frequency or reporting channels. If these reporting systems include malaria data, it should be incorporated into routine malaria data reporting.  
  ● Frequency should be checked and malaria data incorporated into the reporting period in which routine malaria data would normally occur.  
  ● Any incorporated data should specify the source of data (e.g. COVID-19 field hospital).  
  ● If reporting occurs at a health facility, the name of the facility should be incorporated in the data to capture the same spatial resolution as routine data. If specific COVID-19 facilities are performing the tests, their data should be mapped to the nearest health facility.  
  ● All testing, including any that occurs in COVID-19 facilities, should include all routine indicators including signs of fever, number of tests performed, age categories, and test results. |
| **Data quality: divergent data sources**                  | Investigate why the cases do not have a classification, and if the reason is due to mobility restrictions or other disruption, determine whether investigations by phone would be feasible. Explore appropriate responses that may be implemented to prevent outbreaks. |
| **Data quality: Completeness of case investigation and classification** | Outcome: Case management  
Age-specific and urban/rural disaggregation is critical in this context. |
<table>
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| **Testing**          | A low testing or treatment rate may trigger further analysis of the data, including:  
  ● Identify PoCs with low testing or treatment rate.  
  ● Explore whether low testing or treatment aligns with RDT stockouts or lack of access to health care.  

The actions should address the root cause of low testing or low treatment and be targeted to PoCs that drive most of the observed pattern (e.g. those with historically high patient case loads) and have weaker indicators, including:  
  ● Low RDT or treatment availability will trigger similar action at the manufacturer or distribution level. It may include restocking PoCs with commodities that are available at provincial/central level.  
  ● If there is a low uptake of RDTs or treatment (while commodities are available) sensitization and/or training (virtual) could be delivered to health staff on continuing proper case management. The low uptake may also be due to lack of personal protective equipment (PPE) for health care workers, which would prevent safe case management practices; in this case, PPE should be procured (if possible) and training provided on best practice for its use.  
  ● If there is an issue with reporting of these indicators (which may explain any decrease), actions similar to the ones for the reporting indicators could be implemented.  

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**Impact: Malaria trends**

Age-specific and urban/rural disaggregation is critical in this context.

| Malaria incidence | An increase or decrease in malaria trends (incidence or test positivity rate) may have different causes and there may be multiple factors explaining the observed trends (Table 1). Each contributing factor and associated indicators (Steps 1 and 2: commodity availability, uptake of health care services, reporting, changes in case management practices, disruption in preventive malaria services) should be analyzed against the COVID-19 context (Step 3) to identify and address the root cause. For example, there may be disruption in distribution of nets, which may lead to an increase in malaria cases. Disruptions to vector control in eliminating countries might lead to outbreaks. Similarly, a lack of access to treatment may lead to an increase in the number of cases. |
| Test positivity rate | The WHO provides guidance for the implementation of each of the core malaria interventions (vector control, case management, chemoprevention, and extraordinary measures such as mass drug administration) in *Tailoring Malaria Interventions in the COVID-19 Response*. The malaria program may set up a task force to gather additional evidence and decide on the appropriate response; any interventions implemented to alleviate COVID-19 disruptions should be documented. |
| Trends in febrile illnesses | Any changes to this indicator should be communicated to the COVID-19 response team, and the appropriate response should be implemented according to country-specific COVID-19 response plans. |
| Trends in total hospitalizations | Any changes to this indicator should be communicated to the COVID-19 response team, and the appropriate response should be implemented to strengthen resources in affected areas and investigate etiology of illnesses according to country-specific response plans. |
Acknowledgements

This document was initiated by a taskforce on Routine Malaria Data during the COVID-19 Pandemic as part of the RBM Partnership to End Malaria’s Monitoring and Evaluation Reference Group. It was done in collaboration with the WHO COVID-19 – Malaria Workstream on Modelling, Surveillance and Clinical Epidemiology. The following taskforce members contributed as authors of this document: Abdisalan Noor (WHO), Arantxa Roca-Feltrer (MERG-co-Chair, Malaria Consortium), Arnaud Le Menach (Clinton Health Access Initiative), Chris Lourenço (PSI), John Painter (PMI/CDC), Jon Cox (The Bill and Melinda Gates Foundation), Justin Cohen (Clinton Health Access Initiative), Kimberly Lindblade (WHO), Larry Slutsker (PATH), Lia Florey (PMI/USAID), Nicholas Oliphant (The Global Fund), Médoune Ndiop (MERG-co-Chair, NMCP/Senegal), Misun Choi (PMI/USAID), Molly Robertson (PATH), Hannah Slater (PATH), Yazoumé Yé (PMI Measure Malaria, ICF). The taskforce thanks all other individuals who contributed at various stages in the development of this document and supporting materials. The authors are also grateful for those who provided visualization support: Sarah Burnett (PATH), Christelle Gogue (PATH), Wendy Inouye (PATH). Special thanks also go to Peder Digre (PATH) and Kyra Arnett (PATH) for managing the process, creating templates, and editing and proofreading multiple versions of this document.