

Combining Rapid Antigen Testing and Syndromic Surveillance Improves Community-Based COVID-19 Detection in Low-to-Middle-Income Countries

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2 Surveillance Improves Community-Based COVID-19
3 Detection in Low-to-Middle-Income Countries

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38 **1. Abstract**

39 Diagnostics for COVID-19 detection are limited in many settings. Syndromic
40 surveillance is often the only means to identify cases, but lacks specificity.
41 Rapid antigen testing is inexpensive and easy-to-deploy but concerns remain
42 about sensitivity. We examine how combining these approaches can improve
43 surveillance for guiding interventions in low-income communities in Dhaka,
44 Bangladesh. Rapid-antigen-tests and PCR validation was performed on 1172
45 symptomatically-identified individuals at home. Statistical models were fit to
46 predict PCR status using rapid-antigen-test results, syndromic data, and their
47 combination. Model predictive and classification performance was examined
48 under contrasting epidemiological scenarios to evaluate their potential for im-
49 proving diagnoses. Models combining rapid-antigen-test and syndromic data
50 yielded equal-to-better performance to rapid-antigen-test-only models across
51 all scenarios. These results show that drawing on complementary strengths
52 across two rapid diagnostics, improves COVID-19 detection, and reduces false-
53 positive and -negative diagnoses to match local requirements; improvements
54 achievable without additional expense, or changes for patients or practition-
55 ers.

56 2. Introduction

57 Identification and isolation of COVID-19 cases remains key to the pandemic
58 response. The faster and more accurately cases can be identified, the more
59 effectively clinical care can be provided, and transmission reduced through
60 targeted interventions. Real-time PCR has rapidly become the gold-standard
61 test for SARS-CoV-2 detection (although see Dramé *et al*)^[1] due to its high
62 sensitivity and specificity.^[2] However, turnaround can be slow and access to
63 laboratory diagnostics is limited in many parts of the world. As such, syn-
64 dromic surveillance has often been the primary means of case identification
65 for guiding individual and population-wide mitigation measures.^[3,4] Rapid
66 antigen tests are an increasingly popular alternative to PCR as they have
67 high specificity, and are less expensive, easier to perform, and faster, return-
68 ing results within 20 minutes. Hence, rapid antigen tests have potential to
69 greatly decrease the time and expense associated with case detection, but con-
70 cerns have been raised that their lower sensitivity leads to unacceptably high
71 false negative diagnoses.^[5-8] Improving COVID-19 diagnosis is therefore a
72 priority and requires us to better harness imperfect but fast and inexpensive
73 methods.^[9]

74 Syndromic surveillance has been used since the start of the pandemic.^[10]
75 The COVID-19 case definition was based on early data from clinical cases,^[11]
76 but, as the virus has evolved and spread, the clinical picture of COVID-19 has
77 changed. Updated case definitions have improved, though are necessarily non-
78 specific and generate many false positive diagnoses (and ignores asymptomatic
79 cases entirely).^[12,13] A natural extension is syndromic modelling, whereby
80 symptomatic and risk factor data are used to fit a model to allow more ac-
81 curate prediction of how likely a patient is to have COVID-19.^[14] However,
82 disease syndromes change between populations, when new variants emerge,
83 and as other diseases become more or less common,^[12,15] which can make syn-
84 dromic models perform poorly in new settings across space and time. This
85 is a particular challenge for seasonal respiratory pathogens, where symptoms
86 often co-occur and are non-specific.^[12]

87 A key limitation of both rapid tests and syndromic surveillance is their low
88 effectiveness at COVID-19 detection in asymptomatic patients. Asymptomatic
89 cases are known to play a role in driving transmission.^[16] Unfortunately, re-
90 sources are so limited in low- and middle-income settings that health agencies
91 and governments have resolved to focus exclusively on symptomatic patients
92 for both provision of care and intervention to reduce transmission due to their
93 larger contribution to transmission. Asymptomatic cases can still be identi-
94 fied through contact tracing from symptomatic patients. Reliable diagnosis
95 of symptomatic cases of COVID-19, therefore, is essential in low- and middle-
96 income countries.

97 Even for symptomatic patients, neither rapid tests nor syndromic surveillance

98 can match PCR in terms of both sensitivity and specificity. However, lower
99 sensitivity and specificity may be admissible depending on the scale and im-
100 pact of misclassification.^[17] Low specificity means more common COVID-19
101 misdiagnoses (false positives), leading to unnecessary self-isolation, which
102 is expensive to individuals and society.^[18] Low sensitivity means COVID-
103 19 cases will be missed (false negatives) and mitigation measures not put in
104 place.^[19] These misclassifications are complementary for a given diagnostic,
105 meaning increasing specificity will lead to decreased sensitivity, and vice versa.
106 The typical approach is to maximise the number of correct classifications
107 and assume that both misclassification types are equally costly. But, if the
108 disease is prevalent or increasing, false negatives will have an outsized and
109 costly impact.^[19] Or, under low prevalence, false negatives will be correspond-
110 ingly low so even a high false negative rate (low sensitivity) will have modest
111 impact, but small decreases in specificity will lead to a large number of expen-
112 sive false positives.^[20] In practice the situation will be more nuanced and mod-
113 ulated by testing capacity constraints, requiring a balance to be struck.^[17]

114 The best diagnostic approach for surveillance will therefore be one where
115 correct classifications have highest value and misclassifications have lowest
116 cost. Here, we examine the use of rapid antigen testing and syndromic surveil-
117 lance of COVID-19 in symptomatic patients from low-income communities in
118 Dhaka, Bangladesh, where a large volunteer workforce supports COVID-19
119 diagnosis, care and prevention. We demonstrate that by combining rapid anti-
120 gen testing and syndromic surveillance we can draw on their complementary
121 strengths, ameliorate their respective weaknesses, and tune them for differ-
122 ent epidemiological scenarios. We compare their performance alone and in
123 combination for general prediction and as diagnostics under three scenarios
124 with different misclassification requirements. Overall, we show that the op-
125 timised combined models achieve equal-to-much-lower error rates than the
126 next best method in all metrics, and how integrating data from multiple rapid
127 testing methods can improve diagnostics, particularly when adapted to local
128 situations.

129 **3. Results**

130 Of 1241 participants enrolled by community support teams across Dhaka,
131 1172 had complete data available for analyses. The remainder were removed
132 due to duplicated sample identification codes that prevented reliable matching
133 of test results to symptom metadata. These duplications occur at random,
134 due to human error, and we do not believe they could bias results. Patient
135 summaries by age, gender, case positivity and symptoms are presented in Ta-
136 ble 1. Case positivity in Dhaka increased from 15.8% to 23.8% from the first
137 (19th-26th May) to the last week (4th-11th July) of the study, corresponding
138 to prevalence rising from 1.4 to 13.8 confirmed cases per 100 000 people.

Table 1: Breakdown of patient numbers by age and gender, in relation to case positivity by PCR and reported symptoms (both as % rounded to nearest integer). Although age is binned here, raw age in years was used for analyses. Furthermore, in the survey non-binary genders were permitted but none reported.

Age (years)	Gender	Count	Positivity Rate (%)	Symptoms (%)													
				Breathing Problems	Cough	Diarrhoea	Fever	Headache	Loss of Smell	Loss of Taste	Muscle Pain	Red Eyes	Runny Nose	Sore Throat	Tiredness	Vomiting	Wet Cough
16-25	Women	133	20	24	71	4	94	77	39	51	53	11	50	44	74	20	20
16-25	Men	168	20	20	71	5	91	74	45	48	49	9	36	42	62	12	21
26-35	Women	167	25	27	71	10	90	74	37	44	50	4	40	42	68	11	19
26-35	Men	194	27	26	77	10	88	73	39	39	50	7	37	31	70	15	14
36-45	Women	111	28	27	77	4	94	77	40	50	55	5	45	41	74	18	24
36-45	Men	128	26	22	70	7	90	72	38	39	56	9	40	41	68	8	17
46-55	Women	74	22	18	72	3	88	73	34	35	53	0	35	35	58	14	14
46-55	Men	64	25	14	59	5	86	59	33	33	53	11	45	31	70	8	16
56+	Women	64	22	22	70	9	84	56	36	30	45	3	33	27	61	12	22
56+	Men	69	25	30	65	4	77	59	41	36	46	7	38	23	54	13	16
All		1172	24	23	71	6	89	72	39	42	51	7	40	37	67	13	18

139 Model selection for both Model Classes 2 (syndromic-data only) and 3 (syn-
140 dromic and rapid-antigen-test data) showed a marked decline in predictive
141 power at more than 4 symptoms. The final four symptoms retained in Model
142 Class 2 were fever, diarrhoea, vomit and loss of taste and in Model Class 3
143 were loss of taste, dry cough, wet cough and fever. The symptoms are listed
144 in the order they removed through model selection (i.e. all four symptoms
145 were retained in the four symptom model, the first was removed in the three
146 symptom model, the second was also removed in the two symptom model
147 etc.). The covariate gender was dropped for both model classes while age was
148 dropped in Class 2 but retained in Class 3.

149 In the comparison of model predictive performance, Model Class 1 (rapid-
150 antigen-test only) performed worst with an out-of-sample cross-entropy of
151 3.24 (cross-entropy values further from zero correspond to worse predictive
152 performance). The median cross-entropy values were between 2.53 and 2.59
153 for models in Class 2. Models in Class 3 performed best with cross-entropy
154 values between 1.44 and 1.47 (see Figure 1).

155 Generic model classification performance for the one and four symptom mod-
156 els in Classes 2 and 3 is shown by their ROC curves (Figure 2). The curves
157 for the models of different complexities are extremely similar (as are the two
158 and three symptom model curves, not shown), however, note that the four
159 symptom model has higher precision and granularity across both axes. The
160 Class 1 model is a binary test (rapid-antigen-test positive or negative) and so
161 the ROC is a single value, not a curve.

162 Scenario-specific classification performance is shown in Figure 3. Across all
163 scenarios (defined in Table 2), the best models in Class 3 that used both
164 the rapid antigen testing and syndromic data performed equally well or bet-
165 ter than the other two model classes. In Scenario 1 (“Agnostic”), models in
166 Classes 1 and 3 performed equally well (overlapping posterior interquartile
167 ranges) and distinctly better (no overlap in posterior interquartile range) than
168 models in Class 2 (syndromic-data only). The median errors were 0.47 for
169 models in Class 1 and Class 3 and between 0.87 and 0.9 for models in Class
170 2 (Figure 3). In Scenario 2, the model in Class 1 failed to meet the scenario-
171 requirement. The median errors were between 0.75 and 0.76 for models in
172 Class 2, and 0.44 and 0.49 for models in Class 3 (Figure 3).

173 In Scenario 3 (“Low Incidence”), Model Class 2 (syndromic data) again per-
174 formed worst, and Model Class 3 achieved the lowest error, with Model Class
175 1 falling between the two (closer to Class 3 than 2). The error in Class 1 was
176 0.02 and the median errors ranged from 0.19 to 0.2 for Class 2, and 0.18 to
177 0.2 for Class 3 (Figure 3).

178 The candidate models are chosen as a result of a selection process and per-
179 formed much better than more complex models (i.e. with 5 or more symp-

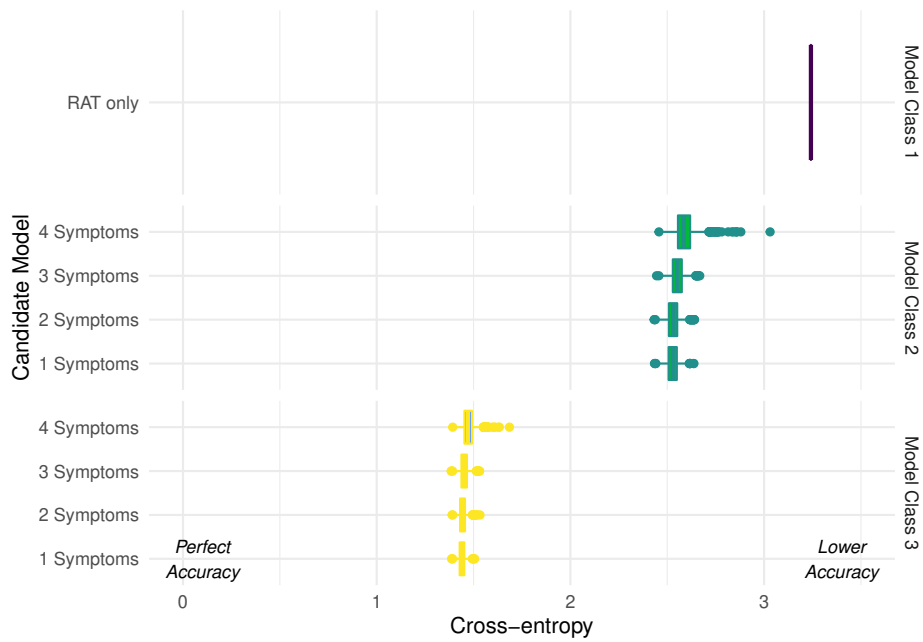


Figure 1: Predictive performance of candidate models. Interquartile ranges for the posterior cross-entropy of the best candidate models at each level of model complexity tested under temporal cross-validation. Values closer to zero indicate better models. The intermediate complexity models perform best at prediction, although performance is similar across all the models within each model class (1. rapid antigen test (RAT) only; 2. syndromic data only; and 3. combined). Models in Class 2 and 3 showed a marked decline in predictive power at more than four symptoms, leading us to choose this as the maximum complexity model in our candidate models.

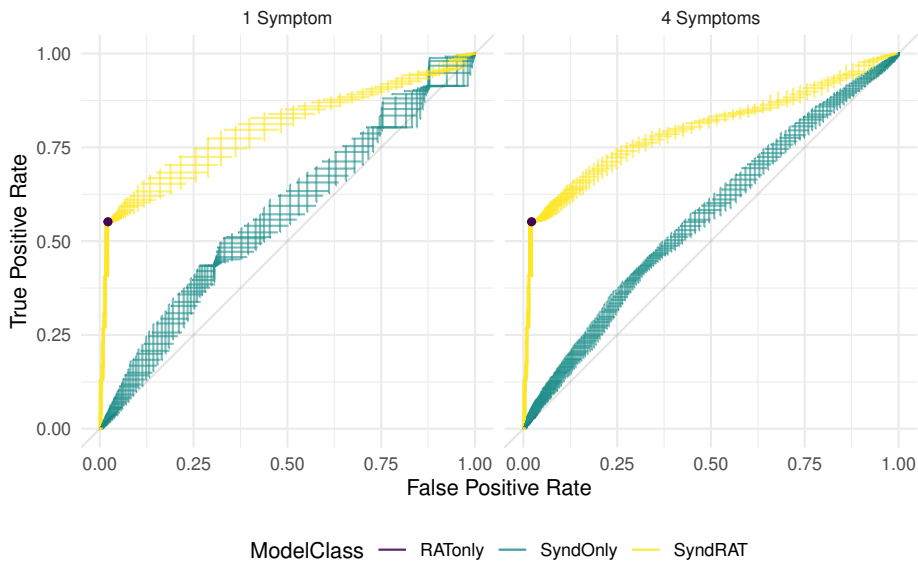


Figure 2: Interquartile ranges for receiver operating characteristics (ROC) for rapid-antigen-testing-only approach (Model Class 1) and posterior median and interquartile range ROC for Class 2 (syndromic-data only) and 3 (syndromic and rapid-antigen-testing data) models. Note that in Model Class 1 the ROC is a single value as the binary test has a single sensitivity and specificity. In Models Class 2 and 3, the ROC are curves which demonstrate the performance of the model for any hypothetical scenario as defined by the axes (as opposed to Figure 5 which demonstrates model performance in specific epidemiological scenarios which are realisations of single points in this space).

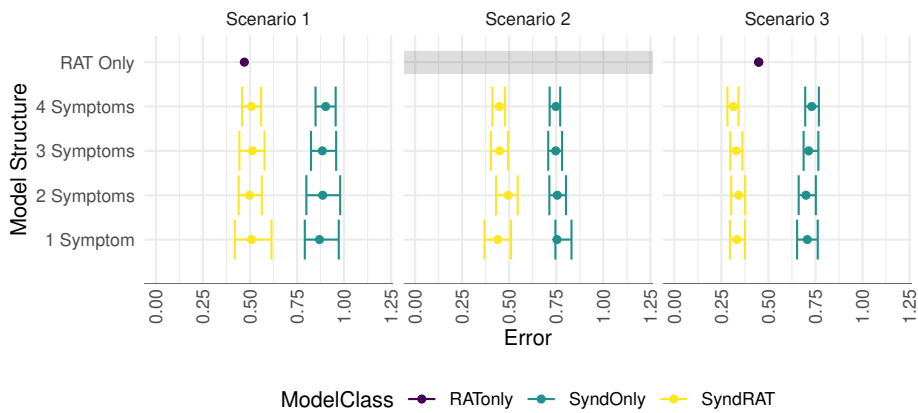


Figure 3: Performance of models under each scenario measured by posterior median and interquartile range for errors defined in Table 1. Lower errors correspond to better model performance. There is no error rate defined for Model Class 1 (rapid-antigen-testing-only model) in Scenario 2 as the model failed to meet the requirement for that scenario (indicated by grey bar).

180 toms) or simpler models (with no symptoms but an intercept and covariates)
181 in terms of cross-entropy and ROC. For Classes 2 and 3 across all scenarios,
182 the number of symptoms made relatively little difference within the final four
183 candidate models in terms of median performance, although the more complex
184 models have higher precision.

185 Across all metrics, the rapid antigen test result is the most informative data-
186 type for potential COVID-19 patients. However, incorporation of even one
187 symptom and the use of a modelling framework greatly improves our ability
188 to predict and classify cases, both generically and in specified scenarios. In-
189 cluding additional symptoms and covariates provides further information on
190 the patient’s status and greater model flexibility, resulting in higher precision
191 in predictions and classifications.

192 **4. Discussion**

193 We have demonstrated that combining rapid antigen tests with syndromic
194 modelling yields better identification of COVID-19 cases than either diagnos-
195 tic in isolation. These gains in performance are mirrored across metrics of
196 prediction, as well as general and scenario-specific classifications. The biggest
197 improvement is seen under the scenario of “Epidemic Growth” (see Table 2)
198 as experienced recently in Bangladesh (time of writing, September 2021), and
199 as expected following relaxation of restrictions and with the emergence of new
200 variants. In this scenario, the combined data model has a false negative rate
201 26 (IQR: 24-29) percentage points lower than the rapid antigen test model.
202 Although the syndromic model matches the combined model’s false negative
203 rate, its false positive rate is 31 (IQR: 29- 34) percentage points higher. Simi-
204 larly, the combined model class performs equally well or better than the other
205 models for the other scenarios explored (Figure 3). These scenarios offer snap-
206 shots of performance, while the model prediction and classification metrics
207 provide an indication of how the models perform more generically (Figures
208 1 and 2, respectively). The more complex model classes are flexible, so can
209 be tailored to specific needs, and benefit synergistically from combining rapid
210 antigen testing with the non-specific syndromic data. Applying our framework
211 to the thousands of cases confirmed daily in Dhaka by PCR, mass deployment
212 of rapid antigen tests with syndromic surveillance can catch tens to thousands
213 of cases that would otherwise be missed.

214 The final symptoms and covariates chosen through model selection should
215 be interpreted cautiously. These models were developed for prediction and
216 classification in a unique sub-population: community support team (CST)-
217 identified, symptomatic patients in low-income communities in Dhaka. Dif-
218 ferent symptoms and risk factors were retained for different model classes,
219 despite data being collected over a short period from the same population.

220 These differences may point to mechanisms by which CST-identified and rapid
221 antigen test positive individuals differ from other groups. They also underline
222 the importance of collecting a relatively broad range of symptom data as the
223 syndromic profile of the disease shifts from population to population. Of interest
224 is whether individuals identified by PCR but missed by rapid antigen tests
225 are less infectious and more typical of asymptomatic cases (perhaps due to
226 different lengths of time since symptom onset). This could be examined using
227 viral load measured as Threshold Cycle (Ct) values from PCR and further
228 testing for other illnesses.^[21] Our use of PCR as a validation test should also
229 be explored further, as it does not have 100% sensitivity so additional validation
230 tests may be informative. It is challenging, however, to find alternative
231 gold-standards that can still be carried out in the community.^[22] Thanks to
232 the modelling framework chosen, it is possible to include additional covariates
233 where they are collected reliably, such as disease prevalence, vaccination rates
234 or the individual's socio-economic status.

235 This boost in diagnostic performance can be achieved with data already being
236 collected in Bangladesh and with methods being rolled out in other low and
237 middle income countries.^[23,24] We ensured our method is scalable by developing
238 it using a large community-based sample and with input from the CST
239 program organisers. As CST data are collected via a mobile phone application
240 the diagnostic model can be updated in real-time. The algorithm of the
241 app could therefore be modified to reflect local epidemiological requirements,
242 informed by local case rates and the considered cost/benefits of misdiagnosis,
243 thereby facilitating adaptation to new variants or even new diseases. However,
244 these models only achieve good performance if validation data are of good
245 quality from the focal population. Similarly, honing your diagnostic to a target
246 misdiagnosis requires an in-depth understanding of local conditions, which
247 can be challenging, requiring socio-economic insight. These are limitations
248 not only of our method, but of all current diagnostic approaches which do not
249 acknowledge them, ignore potential biases and reach only a small (and privileged)
250 proportion of the population. Making these decisions explicit allows
251 them to be more readily challenged, researched and improved upon.

252 Pandemic management can only be done with testing at scale. The combined
253 syndromic and rapid antigen testing approach that we report is promising
254 for large-scale COVID-19 testing in low-income communities. Moreover, our
255 framework is adaptable, including for many other infectious diseases where
256 strict adherence to gold-standard laboratory diagnostics greatly limits testing
257 capacity. Imperfect diagnostics are frequently imperfect in different ways,
258 and these differences are ripe for statistical treatment. These methods are
259 often more agile than gold-standard diagnostics in changing situations as
260 experienced during the pandemic, when fast responses are essential. Overall,
261 our approach shows that by understanding how to utilise the complementary
262 strengths of imperfect but rapid diagnostics (and deploying the more limited
263 gold-standard testing for validation), good quality large-scale testing can be

264 achieved even in low-income communities.

265 **5. Methods**

266 *5.1. Data Collection*

267 Recruitment took place across low-income communities in Dhaka North Com-
268 munity Corporation between 19th May 2021 and 11th July 2021. Participants
269 were identified for COVID-19 testing by community support teams (CSTs).
270 CSTs are community-based volunteer health workers trained to identify indi-
271 viduals reporting symptoms suggestive of COVID-19 through hotline calls or
272 community-based reporting channels. Probable cases identified by CSTs are
273 counselled to isolate for 14-days under household quarantine, connected to
274 telemedicine services for home-based COVID-19 management, and provided
275 with over-the-counter medication or medical referrals if the case is severe.
276 CSTs submit surveillance data to a centralized database through a mobile
277 phone-based application (see Supplementary Materials: Data Collection).

278 Participants were selected for testing if they were over 15 years old, had a
279 fever ($>38^{\circ}\text{C}$), and one or more of 14 symptoms listed in Table 1. CSTs col-
280 lected the enrolled individual’s age and gender, and took two nasal swabs.
281 One swab was used for rapid antigen testing (SD Biosensor STANDARDTM Q
282 COVID-19 Ag Test BioNote) at the household, and the other returned under
283 cold-storage to the Institute of Epidemiology, Disease Control and Research
284 (IEDCR) for PCR testing. The full questionnaire and testing protocols are
285 provided in Supplementary Materials: Data Collection.

286 Participants provided written informed consent to sample collection and
287 for their results to be analysed in the study. The study protocol was ap-
288 proved by the Institutional Review Board at the IEDCR, Ministry of Health,
289 Bangladesh, IEDCR/IRB/04.

290 *5.2. Statistical Modelling*

291 *5.2.1. Structure*

292 We developed three model classes using: 1. the rapid-antigen-test result; 2.
293 the syndromic data, and 3. the two data sources combined (Figure 4). We
294 identified cases by PCR. For Model Class 2, we used a Bayesian multivariate
295 probit model,^[25] with multivariate referring to multiple response variables,
296 not multiple explanatory variables. The multivariate probit structure allows
297 the model to account for the binary and correlated nature of the symptoms,
298 while conditioning on the risk factors of age and gender, thereby improving

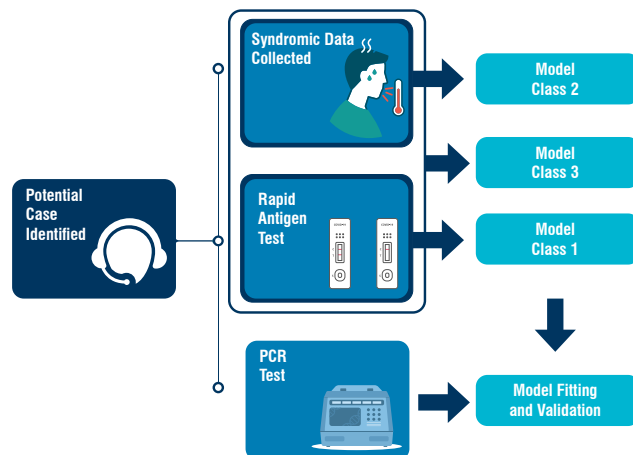


Figure 4: Schematic description of identification of likely COVID-19 cases by community support teams (CSTs) and model definitions. CSTs collect syndromic data (age, gender and presence/absence of 14 predetermined symptoms), and two sets of naso-pharyngeal swabs (for rapid antigen testing and PCR). We used three model classes: rapid-antigen-test-only in 1, syndromic data only in 2, and both rapid-antigen-test and syndromic data in 3. The PCR result is used to train and test each model using temporal cross-validation.

299 over models which implicitly assume independence between symptoms. By
 300 using a Bayesian formulation, we generate full posteriors for our parameter
 301 estimates, allowing natural quantification of uncertainty. For Model Class 3,
 302 we use the specificity of rapid antigen tests by treating rapid test-positives
 303 as cases. While this sounds like a strong assumption, this simply translates
 304 in practice to telling all rapid test-positive individuals to assume they have
 305 COVID-19. Rapid-antigen-test-negative individuals are modelled using the
 306 sensitive syndromic approach of Model Class 2 to capture PCR-positives
 307 missed by the rapid antigen test. This approach leverages the potentially
 308 different syndromic profiles of PCR-positive patients who are rapid-antigen-
 309 test-positive and -negative, allowing the model to adapt solely to the latter.
 310 The models were fitted to the data using Bayesian inference techniques based
 311 on Hamiltonian Monte Carlo in the Stan programming language.^[26] Further
 312 technical details and model equations are presented in Supplementary Materi-
 313 als: Statistical Methodology.

314 5.2.2. Model Selection

315 We conducted backwards model selection, starting with the most complex
 316 biologically plausible model, to identify a subset of models with the highest
 317 predictive power under temporal cross-validation (Figure 5). Shrinking the
 318 number of possible models was necessary to lower computational demand and

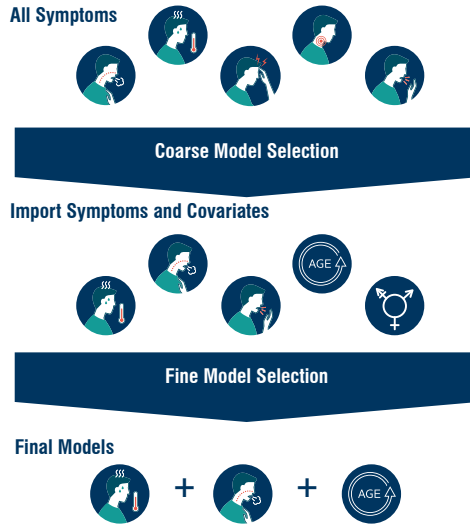


Figure 5: Rounds of model selection in the multivariate probit component of Model Classes 2 and 3. With 14 symptoms (5 shown for demonstration purposes) and two covariates there are over 131 000 possible model combinations. To make exploring these models computationally feasible and to reduce the risk of overfitting, we carried out two rounds of model selection. A subset of symptoms are identified using relationship between each symptom and PCR-status identified by the corresponding model. From this subset of symptoms, a more exhaustive search of potential models is then conducted to identify the best symptom-covariate relationships, using temporal-cross validation to measure model performance. The best model for each level of complexity (i.e. number of symptoms) are then used as our candidate models. Only these final models are used for classification. This reduces the set of models tested as classifiers from >131 000 to just four per model class.

319 reduce the risk of overfitting. The large number of symptoms corresponds
 320 to many potential model configurations (>131 000 for 14 symptoms and two
 321 covariates) which might perform well on the test sets by chance (even under
 322 temporal cross-validation) but lack transferability to novel situations. By
 323 first using the strength of the relationship with the PCR-status (coarse selection,
 324 Figure 5) and general predictive power (fine selection, Figure 5) to
 325 narrow down the number of candidate models, and then testing those models
 326 under the epidemiological scenarios, we are more likely to choose models
 327 that generalise well to new data (see Supplementary Materials: Statistical
 328 Methodology.).

Table 2: Requirements and performance criteria for each epidemiological scenario. The requirement refers to a base level of performance the model must achieve. These requirements were determined through discussion amongst the authors and colleagues at IEDCR. The performance criterion is used to determine which model performs the 'best' given that the requirement has been met.

Scenario Name	Requirement	Performance Criterion (Error)
1 Agnostic	Maximise correct classification rates	Sum of error rates
2 Epidemic Growth	<20% false negative rate	False positive rate
3 Low Incidence	<20% false positive rate	False negative rate

329 5.2.3. Measuring Model Performance

330 We assessed models using three sets of increasingly policy-relevant criteria.
 331 First, we use predictive performance to measure model performance in a
 332 decision-free context (i.e. comparing predicted probabilities of an individ-
 333 ual having COVID-19 to their true status). Second, we use receiver operating
 334 characteristic (ROC) curves to show generic model classification performance.
 335 Finally, we measure classification performance under three epidemiological
 336 scenarios (defined in Table 2).

337 We scored the models' predictive power using cross-entropy (defined in Sup-
 338 plementary Materials: Statistical Methodology). Cross-entropy measures the
 339 accuracy of predicted probabilities of binary outcomes, rather than making
 340 binary classifications, similar in concept to a mean square error for normally-
 341 distributed data, but adapted for binary data.^[27] A cross-entropy value close
 342 to zero corresponds to high levels of accuracy and larger values indicating
 343 lower accuracy.

344 In practice, models are often evaluated on their performance as deterministic
 345 classifiers rather than as stochastic prediction engines (i.e. their ability to clas-
 346 sify an individual as a COVID-19 case or not, rather than the probability that
 347 the individual is a case). Deterministic classification requires that a probab-
 348 ility threshold is chosen over which patients are classified as COVID-19 positive.
 349 Classifier performance was compared generically (using ROC curves to look
 350 at the error rates that can be achieved with each model without specifying a
 351 scenario). Generic performance here is only used to show the flexibility of the
 352 model classes, i.e. model performance without reference to a specific scenario.
 353 The best model for a local situation can only be determined if the relative
 354 costs of false positives and negatives are considered.

355 We compare model performance under three scenarios (using error terms de-

356 scribed in Table 2) developed for illustrative purposes through discussion with
357 colleagues at IEDCR. In Scenario 1, we do not consider epidemiological con-
358 text but minimise false negative and false positive rates equally by maximis-
359 ing the correct classification rates individually and in total, as measured by
360 the harmonic mean (not the arithmetic mean which would maximise the rates
361 in total, see Supplementary Materials: Statistical Methodology.). Scenario 2
362 corresponds to epidemic growth as experienced during the spread of the Delta
363 variant during the period of data collection. Under these circumstances, false
364 negatives are costly relative to false positives. In Scenario 3, incidence is as-
365 sumed to be low and relatively stable. In this situation, policy-makers may
366 prioritize keeping false positive diagnoses low to prevent fatigue and to keep
367 the workforce active.

368 *5.3. Role of the Funding Source*

369 The study funders did not participate in the design of the study, data collec-
370 tion/analysis/interpretation, or in the writing of this paper.

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384 **8. Data Availability**

385 The data and statistical code used in this study are available in a GitHub
386 repository at https://github.com/fergusjchadwick/COVID19_SyndromicRAT_public.

387 **9. Declaration of Interests**

388 The authors declare no competing interests.

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446 11. Supplementary Materials: Statistical Methodology

447 Below we have extended the modelling description provided in the main text
448 to include more technical detail. The code used to implement these tasks is
449 available at https://github.com/fergusjchadwick/COVID19_SyndromicRAT_public.

450 11.1. Modelling

451 11.1.1. Structure

452 We examined the ability of the two imperfect identification methods, syn-
453 dromic modelling and rapid antigen testing (RAT), to predict the patient’s
454 COVID-19 status when used separately and together. These combinations
455 define three model classes (Main Text Figure 4).

456 Model Class 1 uses only the RAT result. It equates being RAT-positive with
457 the patient being PCR-positive for COVID-19 (hereafter, PCR-positive), and
458 being RAT-negative with PCR-negativity.

459 Model Class 2 uses only the syndromic data. For this model, we used a Bayesian
460 multivariate probit model.^[25] The multivariate probit structures the outcomes
461 of the PCR test and symptoms presence/absence as a D -dimensional vector of
462 binary outcomes ($\mathbf{y}_i = (y_{i1}, y_{i2}, \dots, y_{iD}), y_{ij} \in \{0, 1\}$). These outcomes are de-
463 termined by an indicator function which takes a D -dimensional vector of *con-*
464 *tinuous latent* variables ($\mathbf{z}_i = (z_{i1}, z_{i2}, \dots, z_{iD}), z_{ij} \in \mathbb{R}$). These latent con-
465 tinuous variables then covary as realisations of a D -dimensional multivariate
466 normal, with the mean of the error structure informed by a linear predictor
467 (in our case formed of the covariates age and gender), $\sum_{j=1}^J x_{ij}\beta_{jd} + \epsilon_{id}$, and
468 a covariance (Σ) between dimensions. The linear predictor allows us to condi-
469 tion the outcomes on risk factor variables (here, age and gender). The covari-
470 ance structure allows us to account for the correlated nature of the symptoms
471 with each other and the outcome. This multivariate approach (multiple re-
472 sponse variables) is also a very efficient way of encoding complex relationships
473 between symptoms. These relationships need to be accounted for because
474 symptoms are not simply additive in their predictive power. For example, in
475 the diagnosis of measles the “Three C’s” are used: cough, coryza (irritation
476 and inflammation of the mucous membrane in the nose leading to head cold,
477 fever, sneezing) and conjunctivitis. These symptoms individually, and in pair-
478 wise combination could be indicative of a wide range of diseases, but when
479 all three are present measles is a highly probable cause (obviously, this is a
480 simplified example conditioning on patient age and vaccination status). In the
481 alternative, univariate approach, symptoms would be encoded as covariates in
482 the linear predictor for PCR-status, and the complex relationships would need
483 to be reflected as high-order interaction terms. These interaction terms use a

484 large number of parameters and can be hard to fit to data. Using a multivariate
 485 structure allows us to exploit more efficient posterior sampling algorithms,
 486 and in higher dimensional settings like this uses fewer parameters.

487 The covariance matrix formulation of the model described above is not identifiable,
 488 because the variance, $diag(\Sigma)$ and means of the latent variables, \mathbf{z}_i
 489 trade off against each other.^[25] For this reason, we use a correlation matrix,
 490 Ω , formulation with the variance set to 1. A correlation based framework also
 491 makes communication with clinicians and other practitioners smoother as
 492 correlations are more familiar. We thus formulate the multivariate probit as:

$$\begin{aligned}
 y_{id} &= \mathbb{I}(z_{id} > 0) \\
 \mathbf{z}_i &= \mathbf{x}_i \boldsymbol{\beta} + \boldsymbol{\epsilon}_i \\
 z_{id} &= \sum_{j=1}^J x_{ij} \beta_{jd} + \epsilon_{id} \\
 \boldsymbol{\epsilon}_i &\sim N(\mathbf{0}, \boldsymbol{\Omega}) \\
 \Omega_{ii} &= 1 \\
 \beta &\sim N(0, 1) \\
 \boldsymbol{\Omega} &\sim \text{LKJ}(1)
 \end{aligned} \tag{1}$$

493 Model Class 3 combines the two data sources. We utilise the specificity of
 494 RAT by treating RAT-positive patients as PCR-positive patients. The RAT-
 495 negative patients are modelled using the sensitive syndromic approach using
 496 Model Class 2 to capture PCR-positive patients that are missed by the RAT.
 497 This approach leverages the potential different syndromic profiles of PCR-
 498 positive patients who are RAT-positive and -negative, allowing the model to
 499 adapt solely to the latter. Structurally, the model combines Model Class 1
 500 and Model Class 2, with RAT-positive patients being modelled using Model
 501 Class 1, and RAT-negative patients with Model Class 2.

502 By using a Bayesian formulation, we generate full posteriors for our parameter
 503 estimates, allowing natural quantification of uncertainty. Bayesian methods
 504 also facilitate the use of more informative priors. We used minimally informa-
 505 tive priors here. For covariate coefficients (betas) we used standard normals
 506 which are relatively flat in the probit scale. For the correlation prior, we used
 507 the Lewandowski-Kurowicka-Joe (LKJ) distribution, a covariance matrix prior
 508 with unit variance (i.e. a prior for correlation matrices). The LKJ distribution
 509 has a single parameter, η , which controls the degree of marginal correlation
 510 shrinkage. We used minimal shrinkage, $\eta = 1$ ^[28]. More informative priors
 511 that incorporate spatio-temporal effects, for instance, would be natural exten-
 512 sions. The models were fitted to the data using Bayesian inference techniques
 513 based on Hamiltonian Monte Carlo in the Stan programming language^[26].
 514 The models all converged with zero divergent transitions and large effective

515 sample sizes.

516 *11.1.2. Model Selection*

517 We conducted backwards model selection (starting with the most complex,
518 biologically plausible model) to identify a subset of models with the highest
519 predictive power under temporal cross-validation (Main Text Figure 5). For
520 the cross-validation, we divided the data into 5 folds of equal sizes in time
521 order (i.e. the first fold is formed of the chronologically first $\frac{N}{K}$ patients, where
522 N is the number of patients and K is the number of folds, the second fold by
523 the next $\frac{N}{K}$ etc.) To test the sensitivity of this cross-validation structure, we
524 also did a strict temporal division (i.e. the first $\frac{T}{K}$ days where T is the num-
525 ber of days samples were taken on). The results did not change qualitatively
526 between these approaches.

527 The coarse round of model selection (Main Text Figure 5) selected candidate
528 symptoms based on whether they had a strong and consistent correlation
529 with PCR as estimated according to Equation (1). The models were fit with
530 both covariates throughout the coarse round and symptoms were compared
531 in nested models. In the fine round of model selection, these candidate symp-
532 toms and the covariate combinations (age and gender, age, gender and no co-
533 variates) were permuted to more exhaustively explore the model space. Reduc-
534 ing the number of possible models using the two stages of model selection was
535 necessary to reduce computational demand and reduce the risk of overfitting
536 models to the test scenarios. The large number of symptoms corresponds to
537 a high number of potential model configurations ($>131\ 000$ for 14 symptoms
538 and two covariates) which might perform well on the test sets (even under the
539 challenging conditions of temporal cross-validation) but lack transferability.

540 By using general predictive power to narrow down the number of candidate
541 models and then testing those models, we are more likely to choose models
542 that generalise well to new data. It was clear when fitting the models that
543 there were “jumps” in performance (as defined below) between models con-
544 taining five and four symptoms, so the models with one to four symptoms
545 were used as the candidate models. Zero symptom models were not included
546 in the analysis as they do not correspond to a feasible policy (with covariates
547 they would require governments to ask individuals of a given gender and age
548 as COVID-19 positive, and without covariates they would involve randomly
549 assigning individuals as COVID-19 positive).

550 *11.1.3. Predictive Performance*

551 We scored the models’ predictive power using binary cross-entropy (hereafter,
552 cross-entropy). Cross-entropy measures the accuracy of models that gener-
553 ate probabilities of binary outcomes, rather than make binary classifications,

554 similar in concept to a mean square error for normally-distributed data, but
 555 adapted for binary data.^[27] A cross-entropy value close to zero corresponds
 556 to high levels of accuracy, with larger values indicating lower accuracy. More
 557 specifically, the metric allows us to compare a binary vector, $\mathbf{y} \in [0, 1]$, with a
 558 vector of probabilistic predictions ($p(\mathbf{y}) \in (0, 1)$) as follows:

$$\mathbf{H}_p(q) = -\frac{1}{N} \sum_{i=1}^N y_i \cdot \log(p(y_i)) + (1 - y_i) \cdot \log(1 - p(y_i)) \quad (2)$$

559 The resulting score is comparable across all methods for assigning predictions
 560 where the same test data are used, allowing us to compare predictions from
 561 Model Classes 1-3. $H_p(q) \in 0, \mathbf{R}_+$ with zero indicating perfect prediction (as-
 562 signing probabilities of ones and zeroes to outcomes of ones and zeros exactly)
 563 and larger values indicating worse predictions.

564 11.1.4. Classification Performance

565 In applied settings, models must often be evaluated on their performance as
 566 classifiers rather than just as prediction engines (i.e. their ability to say a pa-
 567 tient is COVID-19 positive or negative, not simply the probability the patient
 568 might be COVID-19 positive or negative). To generate a classification, \hat{Y} , a
 569 probability threshold, \hat{p} , must be chosen over which patients are classified as
 570 COVID-19 positive:

$$\hat{Y} = \begin{cases} 1, & \text{if } p(y) \geq \hat{p} \\ 0 & \text{otherwise} \end{cases} \quad (3)$$

571 Receiver operating characteristics (ROCs) are a way to measure the perfor-
 572 mance of a set of classifications in terms of true and false positives and nega-
 573 tives (TP, FP, TN and FN) and the rates of each of these classification types
 574 (e.g. $TPR = \frac{TP}{TP+FN}$, and $FPR = \frac{FP}{FP+TN}$). The error rates are calculated
 575 with respect to a particular threshold, \hat{p} , or across the range of possible \hat{p} s
 576 to generate a ROC curve. In our epidemiological scenarios (outlined below)
 577 we use our ROC curve calculations to identify single thresholds which yield a
 578 required error rate.

579 We strongly emphasise that generic performance here is only used to show
 580 the flexibility of the model classes; the best model for a local situation can
 581 only be determined if the relative cost of false positives and false negatives
 582 is known. Here, we choose three representative scenarios. Each scenario has
 583 a requirement and error rate (defined in Main Text Table 2). We identify
 584 the threshold, \hat{p} , at which the requirement is most closely exceeded (i.e. if

585 the requirement is an error rate should be a maximum 15%, the threshold
586 that produces an error rate below 15% but as close to 15% as possible will be
587 chosen).

588 In Scenario 1, we do not consider epidemiological context but simply minimise
589 false negative and false positive rates equally. We do this by maximising the
590 two correct classification rates both individually and in total, as measured by
591 the harmonic mean. The harmonic mean is used widely in the classification
592 literature as it is maximised by achieving large values in all its component
593 parts, rather than the arithmetic mean which can be maximised by having
594 one extremely large component at the expense of other components. In other
595 words, the arithmetic mean could be large because it has a very high TPR but
596 a small TNR, whereas the harmonic mean will maximise both TPR and TNR.
597 While conceptually the harmonic mean is better suited than the arithmetic for
598 this use case, both produce qualitatively the same results for these data.

599 Scenario 2 corresponds to the situation in Bangladesh at time of writing
600 (September 2021), with COVID-19 cases beginning to rapidly increase again.
601 Under these circumstances, false negatives are extremely costly relative to
602 false positives due to the exponential growth of the disease.

603 In Scenario 3, the pandemic is not declining but maintaining a steady rate
604 of cases. In this situation, policy-makers may be keen to keep false positive
605 diagnoses low to prevent lockdown fatigue and to keep the workforce active.

606 The requirements in Scenario 2 and 3 were developed in discussion with the
607 Institute of Epidemiology, Disease Control and Research (IEDCR), Bangladesh,
608 for illustrative purposes.

⁶⁰⁹ **12. Supplementary Materials: Data Collection**

⁶¹⁰ This document compiles the Community Support Teams' Standard Operating
⁶¹¹ Procedures for the identification of potential COVID-19 patients, screenshots
⁶¹² of the data-collection application, and the protocol for the taking of nasal
⁶¹³ swabs for rapid antigen and PCR testing.

Standard Operating Procedure (SOP) 1 c_for support for Community Support Teams to identify the vulnerable/Risk group

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Preamble:

The Community Support Teams (CST) must be able to follow all the protocols involved in Phase-II of community surveillance of COVID19, in order to control the spread of Coronavirus and find out and protect the vulnerable individuals.

Important Definitions:

Potential Virus Fighter (PVF): An individual reporting symptom of COVID-19. We will be identifying PVFs in the following ways:

- a) through visiting household everyday looking for PVF
- b) through individuals calling government hotlines 333) and reporting any symptom;
- c) other household members of the PVF/VVF who showed symptom being screened

Once the PVF has been screened, the result can be either of these three:

Verified Virus Fighter (VVF): A PVF who has been screened and has high fever and with relevant signs/symptoms of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat) or the loss of the sense of smell.

OR an individual who tested COVID-19 positive in the last few days.

PVF with follow-up: A PVF who has been screened and whose body temperature is between 99.0oF to 99.4oF AND who has at least one sign/symptom of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat, the loss of the sense of smell).

Cleared Virus Fighters (CVFs): A PVF whose body temperature is below 99oF or who does not exhibit any symptoms of respiratory disease (for example cough, shortness of breath, sore throat, the loss of the sense of smell).

Vulnerable Individuals:

Certain individuals are at higher risk of developing complications and dying from COVID-19, these include older individuals (50 years or older in the context of Bangladesh), diabetics, hypertensive individuals, individuals with respiratory diseases such as COPD or those with compromised immune systems. Pregnant women are also a high-risk group for COVID-19 related adverse outcomes.

Scope

For use by CSTs, AMS/VAMs, telemedicine doctors, field Implementation teams and their support teams operating in urban & rural areas of Bangladesh to carry out surveillance for COVID19 and identification and protection of vulnerable groups.

Purpose

The purpose of this Standard Operating Procedure (SOP) is to provide a brief overview of the workflow of CSTs work to a) identify vulnerable individuals efficiently, b) identify the PVF c) take the necessary steps to follow once someone identified as VVF, CVF or follow up PVF. The SOP will link to other technical SOPs and provide guidance to the following activities.

Specific Objectives

1. Reach out to the community and attempt to make contact with households
2. Screen all individuals with COVID-like symptoms (PVFs) to identify if they are VVFs
3. Provide counselling on home management of COVID-19, ensuring 14-day quarantine, mask-wearing guidance to VVFs, and also refer them to CST telemedicine or hospital depending on the severity.
4. Screen all individuals 35 years and older and any pregnant women to identify vulnerable individuals (over 50 years, diabetics, hypertensives, individuals with COPD and pregnant women)
5. Provide counselling on specific protection measures for vulnerable individuals and referral to telemedicine for management of morbidities (diabetics, hypertensive, etc) and pregnancy.

6. Provide SRHR telemedicine numbers to all females aged 15-49 years in the household.

Procedure:

Once the⁶¹⁶ CSTs are trained and grouped into teams and assigned to a particular ward, they will need to do the following coordination activities:

A. Coordination with local authorities Urban:

The AM/VAM (with support from the Field Implementation (FI) team of BRAC will organize in-person meetings for the different wards and zones of the city corporation. The participants should include: the focal person from the ward councilor, the ward councilor, and the zonal executive officer (ZEOs) and the Deputy Chief Health Officer (DCHO). The AM/VAM will support the FI team to inform the local police station about the CSTs working under their jurisdiction. This will include sharing a list of each CST member (along with their photos) working in their particular wards.

Steps in organizing the meeting:

- a) They will be provided the contact details of focal persons and members from Ward councillors.
- b) All physical distancing rules have to be followed: the meeting will only include essential individuals to prevent overcrowding. There should be a minimum of 1-metre distance between each individual.
- c) All participants will perform hand hygiene on arrival and when leaving the meeting and they will all wear masks
- d) Prior to the commencement of the meeting, the meeting venue, including chairs and tables, will be cleaned with disinfectants, especially hard surfaces.
- e) Keep the meeting as brief as possible, try to finish within 20-30 minutes
- f) CST members will check each participant's body temperature before the meeting and maintain hand hygiene throughout the meeting. Participants with whose body temperature is over 99.0oF should be screened as a PVF and cannot join the meeting

Follow-up: On a regular basis, the CST should share activity updates with the local authority /focal person through telephone. The AMs/VAMs will update the FI team regularly, who will also facilitate discussion with the ward councilors/ZEOs/DCHO.

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Follow-up: On a regular basis, share activity updates with the local authority through telephone.

Note: Representatives from the different partner organizations will try to attend the field coordination meetings.

B. Maintenance/Handling of logistics

Refer to SOPS 3 and 5 for materials needed for CSTs to carry out their duties in a safe and professional manner.

CST members will be provided with an Infrared thermometer, oximeter, Wrist watch BP monitor, three-layer cloth mask, measuring tape, gloves, goggles, bleaching powder, disinfectant containing 70% alcohol, id card and vest. They can keep the logistics in house of one of the members. Infrared thermometer, oximeter, three-layer cloth mask will be provided in the training by FAO. Other logistics like more masks and sanitizers will be provided and managed by BRAC.

B. Contacting PVFs

1. Word-of-mouth:

The CSTs will aim to visit an agreed number of households (but the focus should be on complete and comprehensive screening - it is more important to identify VVFs and vulnerable than to maximize household visit numbers). During the visit they will identify PVFs by word of mouth.

2. Government Hotline:

- The CSTs will also have to visit households with PVFs identified through the government hotline. The AM/VAM will contact the PVFs who called in Government hotlines in the last 2-3 days using the phone number used in the call. They will communicate with individuals, guided by a talking point tree, which explains the CST activities and requests the person to allow a screening visit from the CSTs. The contact information will be passed to the specific CST team through the CST mobile app.
- The CST will receive a list of phone numbers of PVFs in their CST mobile app.

- One of the CST members will call the PVF, introduce themselves, describe the purpose of visiting their house and request to schedule a visit. (They will follow the leaflet on FAQ “Coronavirus and CST team related information”).
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- If the PVF is reluctant for a visit, the CST member will try to convince them using interpersonal communication skills. If they still do not agree for the household visit, then the CST member will try to advise them about quarantine over the phone and also ask if they require food or medicine support. The CST will also check if there is any pregnant woman in the house, or any woman who has given birth in the last 6 weeks.

Note: The CST (AM/VAM) should make every effort to contact the PVF, this includes calling each number three times before giving up if it is not answered. If the individual is not willing to have a home visit, the CST will try to counsel them; this may include two calls to try to arrange a home visit.

D. The total Household Visit will include three major activities:

1. Household Form 1
2. PVF Screening (Please follow the [SoP 3 _Quarantine Screening 31 August'20 V8.docx](#) and [SoP 5 _Home Family Quarantine support 310820 V8.docx](#))
3. Vulnerable Screening

Household Form

1. After arriving at a house, the CST will ask to speak to an adult household member and note down the household address in the CST app.
2. The CST will explain about the CST programme, about COVID-19 precautions, and provide CST related leaflet and stickers
3. At this point, the CST will seek consent if the household agrees to a health screening and their information to be passed to health services.
4. Consent Statement: Take consent from the responder. If the household agrees to a health screening and their information to be passed to health services.

5. If the household member doesn't agree then the CST will continue to finalize the form without collecting any phone number and name, give them the CST sticker, thank them and leave the house.

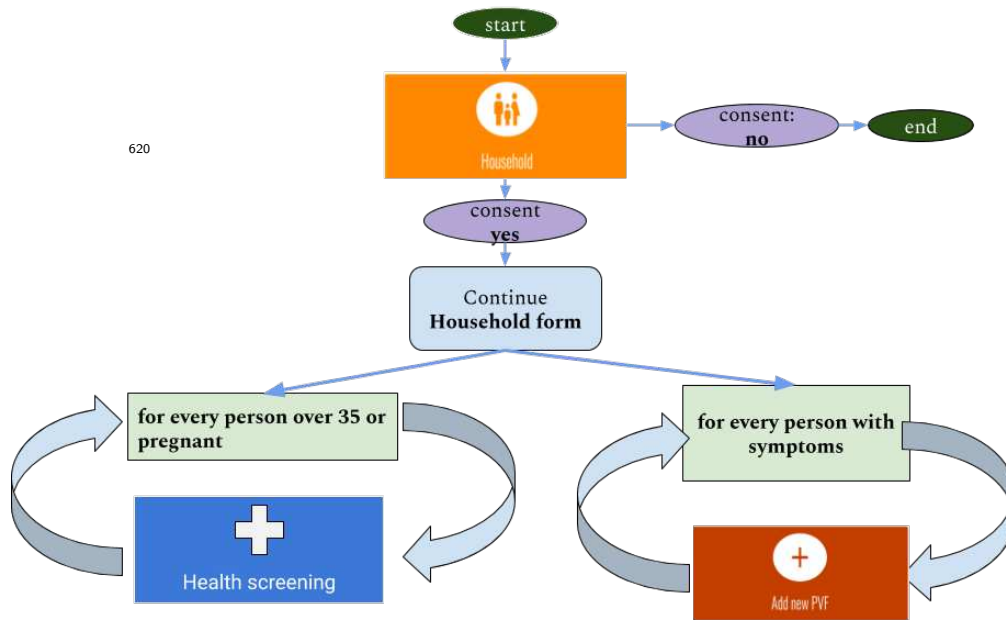
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6. If the household member agrees to continue, the CST will note down then his/her name and phone number as the primary respondent and proceed with the rest of the household form questions (basic questions about vulnerability: Age breakdown of HH members (most important to obtain accurate information about number of individuals over 50 years of age); pregnant, hypertension, diabetes and COPD).

PVF Screening:

7. The CST will then screen each member with COVID-19 symptoms with the PVF screening form will need to be completed for each PVF.
8. At first, the CST will seek consent for PVF screening, if the person doesn't agree then the CST will end the PVF screening. If the person agrees, then the CST will continue with the PVF screening form. This PVF screening process will need to be repeated for each household member with COVID-19 symptoms.

***Consent Statement:** "We will ask simple questions about your health and measure your temperature using an infra-red thermometer. All of your data will be kept confidential under the Ministry of Health and Family Welfare of Bangladesh. Your data might be shared for telemedicine referral and other health-related research or services. You have the right to stop this interview at any point in time or refuse to give answers to any questions that make you uncomfortable.."*



Vulnerable Screening

1. If there is anyone over 35 years of age or if there is a pregnant woman, then please offer them health screening. Please start with the oldest household member.
2. The CST will explain why it is necessary to identify vulnerable people.
3. The CST will follow the App to fill up the first part of the health screening form.
4. Depending on whether the person has one or more declared health issues, the CST will seek consent to do a physical examination to help identify undiagnosed conditions (for example, if an individual says no or does not know about having high blood pressure, the CST will measure their blood pressure).

Consent Statement: *Do you want to have a physical examination now, this will include you measuring your own waist and might include measuring blood pressure.*

5. Once the person agreed to the screening process, please follow SOP 4 c_ on how to measure BP and SOP 4 d_ on how to measure waist circumference.

6. The CST will proceed to ask remaining questions on the screening form.
7. At the end of the screening form, the CST should ask for the respondent's name and phone number.

C. Steps to take for individuals identified as vulnerable:

For individual identified as Elderly, Diabetic, Hypertensive and COPD- provide counselling based on SOP 09 and refer them to CST telemedicine.

Pregnant women: Refer to SRHR telemedicine

D. Steps to take after PVF screening based on the screening result

1. Steps to be taken if PVF is identified as VVF:

- Please refer to SOP 5 Home Family Quarantine Support". Support the VVF in maintaining quarantine for 14-days along with their entire households.
- Measure oxygen saturation levels and enter the level into the App. Take appropriate action based on the oxygen saturation levels.
- Connecting to the telemedicine doctor who will determine the severity of the VVF's symptoms. Depending on the severity, the doctor will recommend a course of action. (Refer to SOP 7: Dedicated Medical Guidance Call Centre for VVF). If it is not possible to connect to the doctor, the CST can leave the number with the VVF to call later. When connecting the VVF with a Telemedicine doctor CST needs to inform the doctor about this. Doctor will decide the severity of the case.
- Provide essential medicine and food support for low income households or arrange for these things to be procured by friends or neighbors.
 - a. The CST will teach the other members of the household on how to avoid direct contact with VVFs while still supporting and motivating the VVF fight against COVID-19. The CST will ensure that the neighbors will understand the role of the VVF in the fight against COVID 19 and are ready to help them.

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- b. The CST will proceed to include information on all household members as per the app specifications (name, phone number, age, gender and relationship with the VVF). If another household member is showing COVID-19 symptoms and wants to be screened as a PVF, only then the CST will screen them as a PVF, otherwise only the information mentioned above needs to be collected for each household member.
 - c. Carry out scheduled follow up visits to ensure adherence to proper quarantine, check if medicines or food is needed and to check if symptoms have worsened.
- At the 14th day of quarantine period, doctors from telemedicine will call VVF to find out the current status if no further sickness is in the household, they are all free to end isolation.

2. Steps to be taken if the PVF is identified as PVF with follow up

The CST will counsel them about monitoring symptoms very closely, and call the CST right away if l if the symptoms worsen

- a. The CST will share their phone numbers if they need further support and will advise of any follow up visits
- b. In any event, the CSTs will revisit him/her within two days to reassess their symptoms and start the whole screening processes again by following the relevant section in the App.
- c. The App will determine if the person is VVF, PVF with follow up or cleared PVF;
- d. If the app changes the status of the PVF with Follow-up to VVF, then the CST will follow SOP 5 as outlined above.
- e. If the app keeps the status of the PVF with follow up, the CST will ask the PVF to contact them if the symptoms worsen. If PVF doesn't contact the CST, CST does not need to visit the household further.
- f. Cleared PVF, the CST will follow the SOP 5 for these categories.
- g. The CST will advise the PVF with follow-up to call the CST immediately if symptoms worsen. The CST will always also advise the entire household to wear masks when going outside their homes and to request visitors to wear masks when visiting.

- h. If anyone in the family develops cough or fever, they may report again contacting their local CST or using 333 or 16263.

3. Monitoring VVFs and PVFs with follow-up:

- Monitoring visits will clearly schedule and are designed to:
 - a. check the health status of the household
 - b. check for compliance with isolation; this should include problem solving if the family are having trouble access food or medicines
 - c. Ensure that the family are not being subjected to stigmatization from neighbours.

Standard Operating Procedure (SOP) 3 for Screening Potential Virus Fighters

Preamble

Once a PVF is identified and details entered into the CST Mobile App, the job of the CST is to screen the CST for COVID19 as soon as possible. The screening process is assisted by the App, which will confirm the status of the PVF.

Important Definitions:

Potential Virus Fighter (PVF): An individual reporting symptom of COVID-19. We will be identifying PVFs in the following ways:

- a) through word of mouth from the community
- b) through individuals calling government hotlines 333) and reporting any symptom;
- c) other household members of the PVF being screened

Once the PVF has been screened, the result can be either of these three:

- 1. Verified Virus Fighter (VVF):** A PVF who has been screened and has high *fever* and with relevant signs/symptoms of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat) or the loss of the sense of smell.
OR an individual who tested COVID-19 positive in the last few days.
- 2. PVF with follow-up:** A PVF who has been screened and whose body temperature is between 99.0°F to 99.4°F AND who has at least one sign/symptom of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat, the loss of the sense of smell).
- 3. Cleared Virus Fighters (CVFs):** A PVF whose body temperature is below 99°F or who does not exhibit any symptoms of respiratory disease (for example cough, shortness of breath, sore throat, the loss of the sense of smell).

Vulnerable Individuals:

Certain individuals are at higher risk of developing complications and dying from COVID-19, these include older individuals (50 years or older in the context of Bangladesh), diabetics, hypertensive individuals, individuals with respiratory diseases such as COPD or those with compromised immune systems. Pregnant women are also a high-risk group for COVID-19 related adverse outcomes.

Community Support Team (CST): In urban and residential areas, the CST will consist of at least two volunteers from different volunteer organizations (e.g., Platform, CDP, Utshorgo foundation, Young Bangla), students from the communities and/or volunteers nominated by the Ward councilors.⁶²⁵

In urban slum area the CST will consists of two Shasthyo Kormi (SK) from BRAC.

Each CST team will be assigned to one ward, and they will be supervised by Area Managers (AM) or Volunteer Area Managers (VAM).

These AMs/VAMs will be responsible for multiple wards (and hence multiple CSTs).

Scope

The SOP is used to determine the status of PVFs by identifying signs and symptoms of COVID19. The PVFs may be identified by the hotline, the community, or as close contacts of a VVF or identified COVID19 patients.

Purpose

The purpose of this SOP is to provide detailed guidelines to the CST on how to screen **PVFs** to determine if they are

- a) VVFs
- b) PVF with follow-up
- c) Cleared PVF

Logistics required (in necessary quantity as per visit plan) for CST:

- Smart phone/tab
- Soap and clean water or alcohol-based hand rub
- PPE items (mask, goggles, gloves)
- Infrared thermometer
- Oximeter
- Biohazard bag/thick poly bag/covered container
- A bucket of prepared diluted 0.2% sodium hypochlorite/bleach solution
- All materials required to make a diluted 0.2% bleach solution for demonstration if PVF is found to be a VVF

- All necessary supportive medicines (first line treatment advised by government telemedicine number 16263) will be carried for distribution to VVF household.
- Phone numbers of local ME

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Procedure

1. The CSTs will make a daily plan for household (Khana) visits to identify PVF by word of mouth as part of their daily work and the PVFs on the App provided by Area managers.
2. The CST will visit household as per given daily target and look for PVFs to be screened.
 - CST will start the conversation with permission and introduce themselves, explain the HH members why they are here.
 - Then they will ask if someone in the household is sick or want to be screened as a PVF
 - If there is any sick person or the member of household want to be screened, please follow the steps from “6 to 14”
 - If the household denied for screening please follow steps “4 and 5”.
3. The CST calls the number from the app to confirm the name and number of the PVF and to request a screening visit. If the number does not answer, they should try three times before reporting the number as not answering.
4. If the PVF is reluctant for a visit, the CST member will try to convince them using good interpersonal communication skills. If they still do not agree for the household visit/screening, then the CST member will try to advise them about quarantine over the phone and also ask if they require food or medicine support. The CST will also check if there is any pregnant woman in the house, or any woman who has given birth in the last 6 weeks. The CST should not give up on the home visit but call again at another time.
5. The CST should use good communication skills to build trust with the PVF and household members before starting the screening. They should remember to treat the PVF as an equal and to respect his/her concerns. They should explain clearly why they are wearing PPE and why they will be taking measurements and asking questions. They cannot enter the house and commence the screening without permission of the household members.

6. Before entering the house, the CST members should wash their hands using soap and water or an alcohol-based hand rub prior to donning appropriate Personal Protective Equipment (PPE):
 - a. Eye Protection: Goggles
 - b. A clean three layer cloth mask
 - c. A pair of new gloves
7. PVF should collect information in a respectful way and record the data on the CST mobile app. If the app does not work for some reason they will record the data in the given PVF interview form. The order of questioning and data collection will be guided by the app or interview form.
8. All personal data and a complete history of the PVF's symptoms should be recorded on the CST mobile app or the form.
9. The PVF's temperature should be taken with an infrared thermometer and the measurement recorded on the CST mobile app. CST members will point the thermometer in 3 centimeters distance from the forehead of the VVF. Please refer to SOP 4a Using the Infrared thermometer.
10. Based on the signs and symptoms and temperature reading, the mobile app will determine if the PVF meets the definition for a **VVF** or **PVF with follow-up** or a **Cleared PVF**.
11. The CST will also check if there is any pregnant woman, any woman who has given birth in the last 6 weeks (42 days) or any vulnerable people in the household.
12. **If the PVF is a VVF, then the CST will take the following steps:**
 - a) The CST will measure the blood oxygen saturation of the VVF using the pulse oximeter and record the reading in the CST mobile app (see SOP 4b. Using the oximeter).

- b) If the oxygen saturation level is equal to or below 93%, the CST should explain to the VVF that he needs specialized medical treatment and immediately call the AM/VAM⁶²⁸ for assistance to take the VVF to hospital.
- c) The CST should proceed to ask the rest of the questions as prompted in the CST mobile app.
- c) The CST should add information (name, age, sex, telephone number and relationship to the VVF) of each household member in the CST mobile app.
- d) These household members should be screened as PVFS.
- e) The CST will check if any member of the household (including the PVF who was just screened) is either pregnant or a breastfeeding. The household will be given the OGSB number to call for any advice on referral to a hospital or any other issue. If a female household member is pregnant: The CST will advise them to go to a health facility for regular antenatal visit, and to deliver in their facilities. -If a female household member is a breastfeeding mother: The CST will advise them to wear masks while breastfeeding, and for them to consider family planning.
- f) The CST should connect to the telemedicine doctor and hold the conversation on speaker phone so that both the VVF and CST can hear. The medical expert who will determine the severity of the VVF's symptoms and depending on the severity, will recommend a course of action. (Refer to SOP 9: Dedicated Medical Guidance Call Centre for VVF). If it is not possible to connect to the doctor, the CST can leave the number with the VVF to call later.
- g) The CST should then follow the Home Family Quarantine Support SOP 5 for guidance on counselling the VVF and their family for maintaining 14-day home quarantine, implementing IPC within the household and support measures.

h) The VVF should be advised that there will be personal follow up visits on days 3 and 7 and then a phone check up on day 10.

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i) The CST will then ask the VVF if they want to be tested for COVID-19, if the opportunity arises. Before recording the answer in the CST mobile app, the CST will clearly describe the consequences of agreeing to be tested as described below:

Please tell them that if they opt for testing, then the government (IEDCR) may call them for sample collection. If the family agrees to be contacted by the government for testing, then they might get a call from the government, and then two or more individuals might visit their house to collect the sample(s). These individuals might be wearing coveralls or other PPE items that might scare the neighbors. There is also a chance that the neighbors might ostracize/avoid the family because of this. If the VVF is less than 18 years old, then please ensure that their at least one of their parents/guardians is part of this conversation. The VVF and their family have complete autonomy to decide if they want to get tested or not. Their decision to test or not to test will not impact the services or support that they are supposed to receive from the CSTs.

Please refer to the Sample Collection SOP for details.

13. If the PVF is a PVF with follow-up, then the CST will take the following steps:

- a) Counsel them on monitoring their symptoms and the importance of contacting the CST right away if symptoms worsen. They should advise on wearing masks outside the house for all household members
- b) The CST will revisit the PVF with follow-up in two-days' time for re-screening.

14. If the PVF is a Cleared PVF, then the CST will take the following steps:

- a) ⁶³⁰ Counsel them on monitoring their symptoms and the importance of contacting the CST right away if symptoms worsen. They should advise on wearing masks outside the house for all household members.
 - b) The CST won't revisit the Cleared PVF again unless they call back for further support.
2. At this point, if there is any household member who is also reporting COVID-19 like symptoms and wants to be screened as a PVF, then the CST can screen this person as a PVF (recording this person as a new PVF screening in the CST mobile-app).
 3. At the end of each household screening the CST members should remove the gloves carefully so as not to allow the outer surface of the gloves to contact their skin and then dispose of them appropriately in a covered container.
 4. They should wash their hands using soap or clean with an alcohol-based hand rub before donning a new set of gloves.
 5. Masks should be worn all day replaced whenever they become wet or visibly soiled and disposed appropriately in a thick poly bag/covered container. Refer to SOP 8 for correct donning and doffing of masks and mask care.

Standard Operating Procedure No 5 for Home Family Quarantine/Isolation of Verified Virus

Preamble

The isolation of the Verified Virus Fighter (VVF) and family is to prevent uncontrolled spreading of the virus responsible for COVID19. To ensure that the VVF and family are able to follow the 14 days isolation, he/she should be treated in such a way that he/she is comfortable and willing to remain in isolation. The other members of the household should be protected from COVID19 infection from the VVF through prevention control practices.

Important Definitions:

Potential Virus Fighter (PVF): An individual reporting symptoms of COVID-19. We will be identifying PVFs in the following ways:

- a) through word of mouth from the community
- b) through individuals calling government hotlines 333) and reporting any symptom;
- c) other household members of the PVF being screened

Once the PVF has been screened, the result can be either of these three:

1. Verified Virus Fighter (VVF): A PVF who has been screened and has high *fever* and with relevant signs/symptoms of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat) or the loss of the sense of smell.

OR an individual who tested COVID-19 positive in the last few days.

2. PVF with follow-up: A PVF who has been screened and whose body temperature is between 99.0°F to 99.4°F AND who has at least one sign/symptom of respiratory disease (for example cough, shortness of breath (in last 15 days), sore throat, the loss of the sense of smell).

3. Cleared Virus Fighters (CVFs): A PVF whose body temperature is below 99°F or who does not exhibit any symptoms of respiratory disease (for example cough, shortness of breath, sore throat, the loss of the sense of smell).

Vulnerable Individuals:

Certain individuals are at higher risk of developing complications and dying from COVID-19, these include older individuals (50 years or older in the context of Bangladesh), diabetics, hypertensive individuals, individuals with respiratory diseases such as COPD or those with compromised immune systems. Pregnant women are also a high-risk group for COVID-19 related adverse outcomes.

Scope

For use by Community Support Teams (CSTs) once a PVF has been declared a VVF and recommended to follow 14 days of isolation.

Purpose

The purpose of this SOP is to provide guidance for CST on how to advise and support VVFs on self-isolation and on quarantine of household contacts.

The steps in screening of the PVF is covered in the SOP 3 Quarantine Screening.

Procedure

1. Once the PVF has been confirmed to be a VVF by the CST using the Mobile App, the CST should advise the VVF and family of his/her status and explain the role of the VVF in controlling the spread of COVID19. It is crucial that the CST explain the importance of his/her actions for the community and Bangladesh and gain the agreement of the VVFs and their families. Good communication skills are needed. Key points to be made include:

- The COVID19 virus is very contagious and can be spread through sneezing and coughing and touching contaminated surfaces. But the virus can easily be killed by cleaning and disinfection.
- Most people do not get very sick, but a small group may need to go to hospital.
- By isolating the VVF is preventing spread of the virus and is working to protect his/her community. If the virus spreads uncontrollably, the hospitals will be unable to cope and many people will die (can use the fish pond example).
- Isolation is a selfless act that helps others; the VVF is a hero.
- After 14 days of isolation the VVF should be over the COVID virus; it will also be clear if the other family members have also caught COVID19.
- The CST will support the family through the isolation period.

2. Discuss openly with the VVF any concerns and fears that he/she may have. Reinforce that the CST and ME are going to assist them. Explain clearly the assistance that the VVF can expect: this includes support visits, telemedicine, food bank support, access to basic medicines and hospital transfer if needed,⁶³³
3. VVF should be advised to go immediately into Family Quarantine/Isolate in their home with all household members (persons who live in the same home).
4. The process of isolation and quarantine should be carefully explained to the VVF and family in a supportive and non-threatening manner. Home Family Quarantine/Isolation means the VVF and their entire household have to adhere to the following conditions for the next 14-days:
 - a. They should not leave their home for any reason.
 - b. If, by chance, they do come across other people, they should stay at least 1 meter (3 feet) apart.
 - c. They should not go out to buy food or collect medicine: they can ask the CST to support them in the process, ask someone else to drop off medicine or groceries at their home or order them by phone or online.
 - d. They should not allow any visitors, other than the CST or medical persons, in their home.

Household hygiene

1. It is very important to protect other household members from COVID19. The VVF should strictly adhere to the following to prevent infecting other household members:
2. The VVF should remain isolated in a separate room and stay 1 meter (3 feet) from other members of the house.
3. He/she and must wear mask and also all family members must wear mask when more than one person is in a room.
4. Enough food and drinks should be prepared and delivered to the VVF's room but not handed to Him/her. Empty plates and cups should be placed into a bucket at the end of each meal and removed and washed in hot soapy water.
5. If the family uses common bathroom, specific bathroom times should be scheduled for the VVF and space given for him/her to move to the bathroom and back to the bedroom.

6. The family members should continue to communicate with the VVF and provide them with company and reassurance from a distance of 3 feet while wearing masks.
7. The VVF should not share dishes, drinking glasses, cups, eating utensils, bed linen, clothes or towels⁶³⁴ with the rest of the family.
8. VVFs who are breastfeeding mothers can breastfeed their infants wearing a mask. They should thoroughly wash their hands with soap and water or sanitize their hands with alcohol-based hand rub before breastfeeding.
9. To reduce the spread of infection in the home, the VVF and other household members should do the following:
 - a. Wash their hands with soap and water often, for at least 20 seconds, or use an alcohol-based hand rub when soap and water is not available. The CST members will show them how to correctly wash their hands and show a sample of alcohol based hand rub.
 - b. The VVF should wear a cloth mask that covers the nose and mouth when he/she must be around other people or animals, including pets. The mask is not necessary when the VVF is alone.
 - c. All the household members must wear masks at all times inside the house (except when someone is completely alone).
 - d. All the household members should sneeze and cough inside the mask; for sudden onset of coughing or sneezing when they are not wearing the mask, they should cover their mouth and nose with a tissue or sleeve (not hands), put used tissues in the covered waste bin immediately and wash hands afterwards. If the mask gets soiled by cough or becomes wet it should be changed.
 - e. The responsibility of taking care of VVF should be given to the healthiest family member who is without any comorbidities such as diabetes, hypertension, cancer, heart disease, chronic respiratory disease.
 - f. Surfaces that are touched often (like door handles, bathrooms, kettles, light switches, chair arms) should be cleaned regularly using household cleaning products and disinfected with 0.2% bleach.. Electronic items such as phones should be cleaned with alcohol.

- g. If a caregiver or other person needs to clean and disinfect a sick person's bedroom or bathroom, they should wear a mask and disposable gloves prior to cleaning. They should wait as long as possible after the VVF has used the bathroom before coming in to clean or use the bathroom. The area should be cleaned first with soap and water followed by disinfection with 0.2% bleach. The disinfected area should remain wet for 5 minutes and then excess bleach cleaned up with a clean cloth.
 - h. Wash the cloth mask with warm water and detergent every day or soak in 0.2% dilute bleach (Add 2 teaspoons of bleaching powder to 1 liter of water) for at least 1 minute, rinse with water, and then let air dry in sunlight if possible.
 - i. The house should be cleaned with normal household products, such as water and detergent, followed by a 0.2% disinfectant bleach. CST members will demonstrate how to make a 0.2% bleach solution (see below) and give each family a 250 mg packet of powder bleach to clean the surrounding surfaces around the VVF and the bathroom after use.
 - j. Used tissues and disposable cleaning cloths should be placed in garbage/polythene bags and then put into a second bag and tied securely. The bag should be stored for 3 days before putting it in the outside bin. Other household waste may be disposed of as normal.
 - k. Laundry should be washed in the usual way. Laundry that has been in contact with an ill person can be washed with other people's items but they should not be shaken as this may spread the virus in the air.
10. To stay well while at home, the VVF and any ill household members should:
- a. drink plenty of water to stay hydrated,
 - b. The VVF and family members should take Vitamin C, Vitamin D and zinc,
 - c. take paracetamol to help ease symptoms like fever and malaise,
 - d. stay in touch with family and friends over the phone or on social media, to help avoid feeling low or lonely
 - e. try to keep busy; -try activities such as , reading, online learning and watching films
 - f. do light exercise, if he/she feels well enough

11. If the household includes a vulnerable individual (someone who is 60 years old or over, has a long-term condition, is pregnant or has a weakened immune system), the household should try to move him/her to another house for 14 days.
12. If the vulnerable person must stay in the home, the VVF and the vulnerable individual should try to keep away from each other as much as possible by:
 - a. Keeping 1 meters (3 feet) away from each other,
 - b. Should wear mask all time even at home,
 - c. avoiding using shared spaces, such as kitchens or bathrooms, at the same time ,
 - d. opening windows in shared spaces, if possible, for air circulation,
 - e. cleaning a shared bathroom each time it is used, for example by wiping the touched surfaces with a disinfectant,
 - f. using detergent and warm water when washing dishes and dry everything thoroughly,
 - g. not sharing a bed, if possible,
 - h. not sharing towels, including hand towels.
13. If the VVF or another ill household member needs medical help during Family Quarantine/Isolation, he/she should not go to a clinic, pharmacy or hospital. He/she should stay at home and call the community support team or contact the telemedicine doctor dedicated to VVFs . The CST will have the number for the local ME.
14. The CST team members should follow up physically with the VVF and his/her household members on the 3rd and 7th day. During the visit they will check VVF's temperature, oxygen saturation and confirm if the VVF and his family are maintaining quarantine (please see SOP 1 Process Flow_Urban Areas for details) Data on body temperature, oxygen saturation, adherence to home quarantine, will be recorded through the CST mobile app during the follow-up visits. The CSTs will also follow-up through telephone on the 10th day to make sure VVF and his family maintained home quarantine properly and also to enquire if they require further food support/medical attention.
15. During the follow-up visits, the CSTs should counsel the VVF and the family again on steps 2-11 (to reinstate the importance of maintaining quarantine).
16. The CST will ensure that the neighbors understand the fight and are ready to help them morally and mentally to boost up VVF and the family.

Preparation of disinfectant bleach solution

1. A 0.2% bleach solution should be made fresh every day by the CST. They will need the following equipment:

- 10 litre bucket with a lid
- 20% bleach powder
- 1 teaspoon
- 1 plastic or wooden stirrer
- Measure for 500 ml (ie a water bottle)
- Protective equipment such as Mask, gloves, goggles. Solution must be made in an open environment.

2. Before beginning the dilution, wear the PPE.

3. To make 1 liter of bleach solution, pour 1 liter of water into the bucket. Add two teaspoon of bleach powder and mix with the stirrer. Immediately put the lid on the bucket

4. Experience will show how much is needed for a day's work but do not make more than be safely carried in the bucket. The calculation is easy: for 2 liters of water add 4 teaspoons of bleach; for 5 liters of water use 10 teaspoons of bleach powder

5. At the end of the day, pour out any remaining solution

PVF
Screening

←
Household

Address HH5

Respondent name Aa

Respondent phone number 11112236547

Complete household visit 🍷

←
PVF Form

Does the person agree to a health screening and their information to be passed to health services?

All of your data will be kept confidential under the Ministry of Health and Family Welfare of Bangladesh. Your data might be shared for telemedicine referral and other health-related research or services. You have the right to stop this interview at any point in time or refuse to give answers to any questions that make you uncomfortable.

Yes No

1. Phone number

+88

2. Alternative phone number

+88

3. PVF Name

4. Age of the PVF

5. Gender

Male Female Other

10. PVF's temperature reading

°F

11. Symptoms

Cough

Diarrhoea

Headache

Loss of smell

Loss of taste

Muscle pain

Shortness of breath

Sore throat

Tiredness

Red eyes / Conjunctivitis

Runny nose

Sputum production (Wet cough)

Vomiting

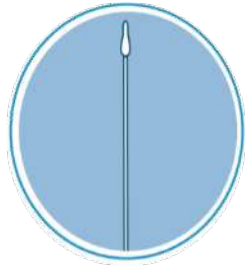
12. Has the PVF tested positive for covid in the last 7 days?

Yes No

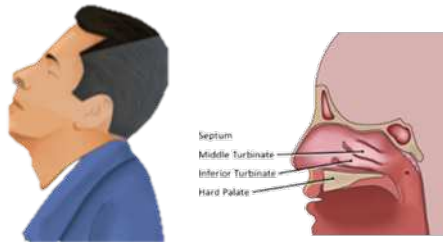
Save and next

Nasal Sample Collection and Testing Protocol

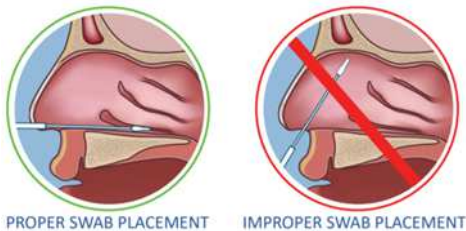
[Nasal sample collection \(infographics showing Left Nasal sample collection\) instruction for CST 1](#)



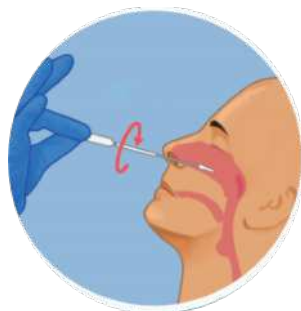
- Take out the nasal mid-turbinate swab from the packet and keep the tube safely for the time being.
- Touch only the plastic shaft not the padded end.



- Ask the patient to sit straight and tilt the head back (approximately 70 degree).

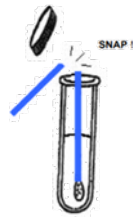


- Insert the swab in the nasal space parallel to the hard palate.
- Resistance will be felt and that is the confirmation of reaching to the nasopharynx.



- Once the swab is against the hard surface rotate it several times.

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- Take out the swab from the left nose and insert the swab into the VTM labelled as "N"
- Make sure the liquid transport medium covers the tip of the swabs.
- Break the swab shafts at the marking on the shaft.



- Screw the caps back on the test tubes tightly.

Once the nasal sample is collected by CST 1, CST 2 will check the box in the app (See example below).

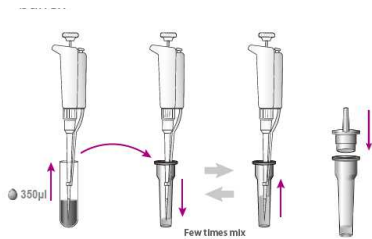
Specimen:

- Collected Not collected

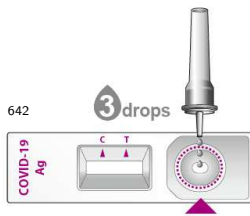
If collected mention type:

- Right nasal swab Throat swab Saliva Combined left nasal swab and throat swab

[Nasal swab sample analysis](#)



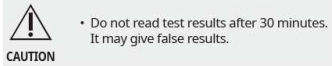
- Using a micropipette, collect the 350µl of specimen from the VTM. Mix the specimen with an extraction buffer in another tube.
- Press the nozzle cap tightly onto the tube.



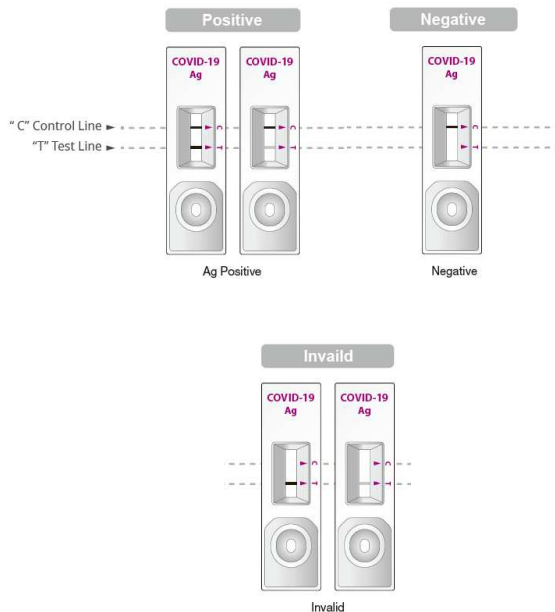
- Apply 3 drops of extracted specimen to the specimen well of the test device.



- Read the test result in 15-30 minutes.



Interpretation of Nasal sample analysis



- A colored band, control line (C), in the top section of the result window will appear in positive and negative test result.
 - Presence of a second colored band, "T" test line, in conjunction with the "C" Control line is always considered as positive. Even if the "T" test line is faint.
 - Presence of only "C" control line without "T" test line will be considered as negative.
- Absence of the control line in the top section will always consider the result as invalid.

Image and Information Sources:

https://www.cdc.gov/coronavirus/2019-ncov/downloads/lab/NMT_Specimen_Collection_Infographic_FINAL_508.pdf

Supplementary Files

This is a list of supplementary files associated with this preprint. Click to download.

- [nrreportingsummaryFJC.pdf](#)