A New Handheld Device for the Detection of Falsified Medicines: Demonstration on Falsified Artemisinin-Based Therapies from the Field

Benjamin K. Wilson,1* Harparkash Kaur,2* Elizabeth Louise Allan,2 Anthony Lozama,1 and David Bell1
1Intellectual Ventures Laboratory, Bellevue, Washington; 2London School of Hygiene and Tropical Medicine, London, United Kingdom

Abstract. Poor-quality medicines are a major problem for health-care systems in resource-poor settings as identifying falsified medicines requires a complex laboratory infrastructure such as a Medicines Quality Control Laboratory. We report here an evaluation of a low-cost, handheld near-infrared spectrometer (NIRS) device by analyzing a library of artemisinin-based combination therapy (ACT) medicines to determine its usefulness as a drug-screening tool. The “SCIo” research prototype device was used to collect NIR spectra of a library of ACT and artemesunate monotherapy medicine samples previously collected in Bioko Island and Equatorial Guinea and Kintampo, Ghana. The quality of these samples had been categorized as falsified, substandard, and quality assured based on the amount of stated active pharmaceutical ingredients detected using high-performance liquid chromatography photodiode array. Numerical analyses were performed on the NIR spectra to assess the usefulness of NIR to identify falsified and substandard medicines. The NIRS device was successful at detecting falsified medicines in all cases where the library contained both quality assured and falsified medicines of the same stated brand of medicines. The NIRS device was successful at identifying substandard amounts of artemesunate but not amodiaquine in the ACT samples (N = 15) of artemesunate–amodiaquine. This work reveals that this low-cost, portable NIRS device is promising for screening ACTs for falsified samples and could enable widespread drug screening at all points of the health system.

INTRODUCTION

Drug quality is fundamental to the success of health interventions, with intentional or inadvertent provision of substandard and falsified medicines being a major contributor to delayed clinical recovery and mortality, particularly due to infectious disease in low-income countries. Use of medicines with subtherapeutic concentrations of the stated active pharmaceutical ingredient (SAPI) may drive microbial resistance. Although awareness of poor-quality medicines, in particular “fakes” has increased appreciably with the publication of reports citing that up to a third of antimalarials have failed chemical content analysis, the ability of health providers and consumers to detect substandard and falsified medication is limited by the high cost, limited accessibility, and destructive nature of testing.

Medicine quality is commonly divided into four classes: quality assured, falsified (also referred to as counterfeit), substandard, and degraded. Quality-assured medicines have the specified amount of SAPI and meet other quality attributes. Falsified medicines do not contain the SAPIs indicated on the packaging and generally do not comply with intellectual property rights. Substandard medicines are produced with poor constituents or manufacturing practice and may have contents or dissolution times that are outside of therapeutic limits or have deteriorated due to adverse environmental conditions. Degraded formulations are commonly the result of exposure to light, heat, and/or humidity. It is clearly important to distinguish degraded from substandard products, as implications for rectifying these conditions are very different. In countries with insecure public or private supply chains, there is a need to detect at the point of importation as well as down the distribution chain to the point of purchase. Arguably, a place for consumer-level testing also exists in more regulated markets, where mail order medication (medicines bought over the internet) is becoming common practice.

Detecting the quality of medicines is a challenge for health-care professionals even though strategies and specifications have been formalized by international organizations such as the U.S. Pharmacopeia, the World Health Organization (WHO), and the Global Fund. Testing predominately requires a well-equipped Medicines Quality Control Laboratory (MQCL) equipped with a range of analytical equipment such as high-performance liquid chromatography photodiode array (HPLC-PDA), mass spectrometry, dissolution testing apparatus, and so on.

Because of the cost and limited availability of MQCL, investigations are generally limited to relatively isolated and high-cost surveys and spot checking by large procurement agencies at or before the level of importation (missing the possibility of later substitution within the supply chain). Hence, there is a need for low-cost, robust, handheld devices for use as screening tools to detect poor-quality medicines with ease throughout the supply chain. Ideally, such a method would also be nondestructive. Devices fitting these criteria could transform pharmaceutical supply lines, providing assurance to health services, local providers, and consumers, while damaging the profitability of the drug counterfeit industry. Nondestructive spectroscopic techniques such as Raman spectroscopy and near-infrared spectroscopy (NIRS) are gaining interest for their ability to test drug samples without using the toxic chemicals or flammable solvents typically used in a MQCL, and with low training requirements. Raman and NIRS are limited to analyzing the state of the surface presented to the detector. Both techniques rely on comparing the characteristic spectral “fingerprint” of a suspect medicine with a quality-assured sample. To distinguish falsified from quality-assured medicines necessitates access to a database of spectra containing the spectral signature of the quality-assured product. NIRS, in particular, requires relatively
complex computer analysis for library comparisons, as spectral features in this wavelength range are less pronounced than those observed by mid-infrared and Raman spectroscopy. Using NIRS to detect subtherapeutic levels of SAPIs necessitates that the specific chemical signature be distinguishable and quantifiable within the spectral pattern, and the surface presented be representative of the quantity within the whole product.

The TruScan® by Thermo Fisher Scientific (Waltham, MA) has been investigated for drug-quality application in field settings and is equipped with software to identify certain artemisinin derivatives. However, adapting the TruScan as well as UV-Vis-NIR devices for the identification of falsified medicines in field settings has been limited. NIRS systems are simpler, requiring only a lamp for illumination and a dispersive optic, and are therefore more amenable to portability and low cost (approximately USD 1,000).

Herein, we evaluate the usefulness of a consumer NIRS technology presently marketed at USD 250 (“SCiO” research model device, Consumer physics, Israel), for its potential to detect poor-quality (substandard and falsified) first-line antimalarial medicines, specifically artemisinin-based combination therapies (ACTs). The ACTs used had been purchased in Bioko Island, Equatorial Guinea, and Kintampo, Ghana, and previously undergone quality testing.

MATERIALS AND METHODS

The NIRS device. The Consumer Physics SCiO research model NIRS device used in this study is a smartphone-sized instrument that promises to be highly effective at checking the quality of medications in the absence of a MQCL. It operates as a diffuse reflectance NIR spectrometer and includes a small, optical integrating attachment similar to an integrating sphere (Figure 1). The device used in this work, a research prototype, was operated using a smartphone app.

Field samples. Artemisinin derivative-containing antimalarials, including ACTs and monotherapies, were purchased in Bioko Island, Equatorial Guinea. A total of 677 samples (77 brands) were purchased using three different systematic sampling approaches (convenience, mystery, and overt) from 278 outlets comprising of pharmacies, patent medicine vendors, public health facilities, and private hospitals (publication under review).

Samples were analyzed for content using the gold standard method of HPLC-PDA to quantify the amount of SAPIs present in each sample (SAPI 1 = artemisinin derivative, SAPI 2 = nonartemisinin antimalarial partner drug). Results were expressed as a percentage of the SAPI indicated on the packaging and samples were classified as quality assured, substandard, degraded, or falsified. The analysis showed that 7.4% of the samples were falsified, 1.6% were substandard, and 91.0% were quality assured. No degraded samples were detected (manuscript for publication in preparation).

Similarly, 257 artemisinin derivative-containing samples (31 brands) were purchased in Kintampo, Ghana, using the mystery client sample collecting approach. Analysis using the HPLC-PDA found 37.0% of the samples were substandard and 63.0% samples were quality assured. Quality-assured samples were so classified when they contained SAPIs within the tolerance band of 85–115% as measured by HPLC-PDA analysis and those out of this range were classified as substandard. The acceptable range for artemisinin-containing antimalarial (ACA) and its companion drugs according to the international pharmacopeia is for it to contain 90.0–110.0% SAPIs. However, the wider tolerance band was selected by the ACTc-DQ team to reflect the typically observed experimental variation in HPLC laboratory testing where less than 10 samples of each ACA formulation were tested. No falsified or degraded samples were detected.

NIRS spectrum acquisition. Sample brands (Table 1) from Bioko Island, Equatorial Guinea, and Kintampo, Ghana, already analyzed for quality using the gold standard technique of HPLC-PDA were used to validate the usefulness of the NIRS device. The HPLC results were not shared at this stage with the persons using the NIR, ensuring that the study was blinded at this stage. Each ACA sample tablet under investigation (677 from Bioko Island and 257 from Kintampo, Ghana) was placed in the integrator attachment of the NIRS and scanned three times. Triplicate scanning reduces signal-to-noise ratio (by averaging scans together) and enables for the consistency of scan quality to be assessed. Bilayer tablets (all of which were Winthrop® (Maphar, Morocco) artesunate–amodiaquine formulations) were scanned three times on each side and the two sides were considered independently for the analysis. The geometry of all the tested bilayer tablets was such that the two layers could be scanned separately using the integrating attachment.

Measurements were carried out in the 740- to 1,060-nm wavelength range, which is the full range of the NIRS device. The acquisition time for a single spectrum was around 2 seconds. The combination of the optical geometry of the integrating attachment and robust spectral preprocessing (described below) allowed spectra obtained with the integrating attachment to be consistent over a wide range of tablet positions within the chamber.
In addition, reference standard compounds of artemether, lumefantrine, artesunate, amodiaquine, dihydroartemisinin, and piperaquine were scanned without the integrating attachment (it is not possible to place them in the integrating attachment without contaminating of the interior surfaces).

Spectral processing. The goal of spectral processing is to remove extraneous signals such as background offsets and noise while enhancing spectral features. Averaged spectra were converted to absorbance units to enable linear fitting with reference standard compound in subsequent regression analysis. Spectra were then smoothed using a Savitzky–Golay filter, a common method that uses a sliding polynomial fit to reduce noise. For qualitative analysis, (identification) the second derivative spectra of absorbance were used to sharpen and enhance spectral features, a technique which produces improved performance in conjunction with library matching algorithms. For regression analysis of API concentration, the smoothed absorbance spectrum was used following a background subtraction.

Analysis to identifying falsified samples. The primary goal of regression analysis was to identify substandard medicines. Herein, three methods were used including a quantitative neural network, principal components regression, and partial least squares regression (PLSR). Of these three, PLSR showed the best results with the least tendency for overfitting. All quantitative results were obtained using leave-one-out cross-validation, or to predict the concentration of each sample a unique calibration curve was produced that did not include data from that sample.

RESULTS AND DISCUSSION

Reference standard compounds versus ACTs. NIR spectral analysis results of samples from Bioko Island were compared with those of the corresponding reference standard compound (Figure 2). Note that formulations of artesunate–amodiaquine are shown as separate spectra despite being part of the same combination therapy, as these compounds (artesunate and amodiaquine salt) are unstable together, which necessitates for them to be formulated either as two separate tablets or as separate layers of a bilayer tablet. Figure 2C shows the combined spectra for both bilayer and multi-tablet therapies.
The spectral signatures shown in Figure 2A represent 12 brands of artemether–lumefantrine, whereas Figure 2C represents 16 brands containing artesunate and eight brands containing amodiaquine (see Table 1). The relatively narrow error bars represented in these plots indicate that the NIR signatures of drugs containing the same SAPI are consistent across different brands, whereas drugs containing different SAPIs have NIR signatures distinct from each other.

Spectral match of NIR peaks of the reference standard compounds and those from the samples of ACTs is shown for artemether–lumefantrine and artesunate (Figure 2A and C). However, this is not the case for peaks of amodiaquine, dihydroartemisinin, and piperaquine and tablets containing these medicines. This implies that a consistent formulation of excipients also contributes to the unique NIR spectral features associated with the SAPIs of the formulation as seen in Figure 2.

Figure 2. Second derivative absorption spectra of different active pharmaceutical ingredient (API) groups matched with the spectra of the associated API standards, when available. (A) Shows artemether-lumefantrine, (B) shows dihydroartemisinin-piperaquine, (C) show artesunate-amodiaquine, and (D) shows artesunate-sulfamethoxypyrazine. Artesunate–amodiaquine is a combination therapy in which the two stated active pharmaceutical ingredients are formulated as either a bilayer or as separate tablets. This allowed their spectra to be collected and presented separately. The spectra in (C) represent scans from both bilayer tablet and multi-tablet doses.

Figure 3. (A) Second derivative near-infrared spectra of quality assured and falsified Coartem. (B) Principal components clustering diagram for the spectra in (A).
Quality-assured formulations of artesunate–sulfamethoxypyrazine–pyrimethamine combination therapy were also tested in the Bioko Island sample set to determine the representative spectrum (Figure 2D). Although there were no falsified samples or reference standards to show spectral comparisons, the pronounced spectral features of this formulation made it easy to identify versus all other formulations.

**Detection of falsified medicines.** The library matching algorithm described in the section analysis to identifying falsified samples was used to demonstrate the ability of a simple classifier to distinguish NIR spectra of quality-assured Coartem® (Novartis, Cambridge, MA) from falsified Coartem samples RL (Figure 3). The NIR spectrum of quality-assured Coartem is reproducible when compared with the falsified samples. This consistency in tablet spectra is most likely due to Coartem being produced by a manufacturer with WHO prequalification status. The Coartem cluster in Figure 3B (blue) contains 23 batches of quality-assured Coartem from two manufacturing locations (United States and China). Unsurprisingly, the library matching algorithm was able to distinguish quality assured from falsified Coartem with complete (100%) accuracy.

Artemisinin-based monotherapy tablets are not recommended for the treatment of malaria by WHO, but were purchased in Bioko Island. The library matching algorithm used was able to distinguish between the falsified and quality-assured monotherapy tablets (Table 2).

**Detection of substandard medicines.** Substandard medicines (i.e., low API concentrations), samples purchased in Kintampo, Ghana, that had been analyzed with HPLC-PDA and classified as substandard were used to assess the ability of NIRS to identify them (Table 3). Regression analysis of the NIRS spectra could not indicate the quantity of amodiaquine in the Arsumoon® [Guilin, China] brand (Figure 4), as the spectra of the reference standard amodiaquine was not distinguishable in the spectral signature of therapies containing amodiaquine (Figure 2). In contrast, changes in the concentration of artesunate in the Combicure Adult brand (where artesunate and amodiaquine are formulated as separate tablets) could be detected, although the accuracy was low ($R^2 = 0.66$). The Combicure Adult sample set had artesunate concentrations determined using HPLC-PDA ranging from 74.8% to 103% of SAPI, corresponding to doses of 74.8–103.0 mg. The quantitation accuracy of this linear model is defined by the standard deviation of the normalized residuals, which was 7.4%. This indicates that the NIRS can quantify the SAPI within ±14.8% with 95% certainty. However, the sample number was relatively low in this regression ($N = 15$) and a larger sample size is needed to determine the significance of this result. Figure 4A and B show the spectra of the two artesunate–amodiaquine brands, Combicure Adult and Arsumoon. Note that the spectra of Combicure Adult indicated systematic variation with SAPI concentration (shown by different colors), whereas the spectra of Arsumoon does not.

### CONCLUSION

Falsified and substandard medicines are a major contributor to morbidity and mortality through treatment failure. The lack of a low-cost, simple, and reliable testing method to assess quality on the ground has restricted the ability of health and enforcement authorities to detect and deal with this issue. The results presented here indicate that handheld NIRS devices offer promise to address at least the first of these problems, that is, detecting falsified products provided that a consistent spectrum from the quality-assured product is available for comparison. NIRS also showed potential for detecting certain medicines with insufficient concentration of the SAPI and hence could be used for identifying substandard medicines, though we were only able to test the Combicure Adult and Arsumoon brands. These results require further validation and indicate that quantitation may not be possible for certain SAPIs, such as amodiaquine. Access to a wider NIR wavelength range of 740–1,060 nm will improve the detection of varying amounts of SAPIs. Handheld NIR spectrometers with wavelength ranges up to 2,400 nm are now available for sale.

The NIRS, coupled with the appropriate spectral library, proved effective for identifying falsified samples of Coartem (artemether–lumefantrine) across multiple batches and manufacturing locations. Based on these results and the underlying technology, it seems likely that handheld NIRS scanners would be highly effective instruments for screening medicines from well-characterized brands with high-quality

### TABLE 2

**Falsified drug identification sample set and results**

| Brand name | SAPIs          | WHO PQ? | Samples | Country of manufacture (no. of batches) | Falsified Samples | Batches | Sens/spec |
|------------|----------------|---------|---------|----------------------------------------|-------------------|---------|-----------|
| Coartem    | Artemether     | Yes     | 60      | United States (15)                     | 22                | 3       | 100/100   |
|            | Lumefantrine   |         |         |                                        |                   |         |           |
| Artesunat  | Artesunate     | No      | 3       | Vietnam (1)                            | 19                | 2       | 100/100   |
| Cusnat     | Artesunate     | No      | 36      | China (2)                              | 1                 | 1       | 100/100   |

API = stated active pharmaceutical ingredients; sens/spec = sensitivity/specificity; WHO PQ = World Health Organization prequalified (manufacturer).

### TABLE 3

**Sample set for determining API concentration**

| Brand          | API group       | Substandard component | API range (%) | Samples | Batches | WHO PQ? |
|----------------|-----------------|-----------------------|---------------|---------|---------|---------|
| Arsuramoon     | Artesunate–amodiaquine | Amodiaquine         | 66–85         | 15      | 7       | Yes     |
| Combicure Adult| Artesunate–amodiaquine | Artesunate          | 75–83         | 10      | 4       | No      |

API = active pharmaceutical ingredient; WHO PQ = World Health Organization prequalified (manufacturer).
manufacturing (Coartem is prequalified by WHO). Falsified samples of other formulations or brands with matched quality-assured samples were only available for artesunate monotherapies.

The development of a library of spectra of quality-assured products approved by a national regulatory authority would allow low-cost rapid testing at port of entry and throughout the supply chain, as well as offer health providers and consumers with a means to confirm the authenticity of the product at point of use. Although technical developments in the near future are likely to enlarge this further, capacity to detect substitution of falsified products into supply chains should now be available to health systems and monitoring services with currently available technology.

Received November 15, 2016. Accepted for publication January 4, 2017.

Acknowledgments: We gratefully acknowledge the Medical Care Development International Bioko Island Malaria Control Project for the purchase of the samples, as well as the teams on the ground, who purchased the samples in Kintampo, Ghana, and Bioko Island, Equatorial Guinea.

Financial support: The authors acknowledge financial support from the Bill & Melinda Gates Foundation, which funded sample collection in Kintampo, Ghana, and high-performance liquid chromatography analysis as part of an award to the artemisinin-based combination therapy consortium at the London School of Hygiene and Tropical Medicine, and Bill & Melinda Gates for their sponsorship of near-infrared spectrometer analysis through the Intellectual Ventures Global Good Fund.

Authors’ addresses: Benjamin K. Wilson, Anthony Lozama, and David Bell, Intellectual Ventures Laboratory, Bellevue, WA, E-mails: bwilson@intven.com, alozama@intven.com, and dbell@intven.com.

Harparkash Kaur and Elizabeth Louise Allan, London School of Hygiene and Tropical Medicine, London, United Kingdom, E-mails: harparkash.kaur@lshtm.ac.uk and louise.allan@gmail.com.

This is an open-access article distributed under the terms of the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.

REFERENCES

1. Kaur H, Clarke S, Lalani M, Phanouvong S, Guerin P, McLoughlin A, Wilson BK, Deats M, Plançon A, Hopkins H, Miranda D, Schellenberg D, 2016. Fake anti-malarials: start with the facts. Malar J 15: 86.

2. Nayyar GML, Breman JG, Newton PN, Herrington J, 2012. Poor-quality antimalarial drugs in southeast Asia and sub-Saharan Africa. Lancet Infect Dis 12: 488–496.

3. World Health Organization, 2011. Survey of the Quality of Selected Antimalarial Medicines Circulating in Six Countries of Sub-Saharan Africa. Geneva, Switzerland: World Health Organization, 1–118.

4. Gautam CS, Utreja A, Singal GL, 2009. Spurious and counterfeit drugs: a growing industry in the developing world. Postgrad Med J 85: 251–256.

5. Attaran A, Berry D, Basheer S, Bate R, Benton D, Chauvin J, Garrett L, Kickbusch I, Kohler JC, Midha K, Newton PN, Nishtar S, Orhii P, McKee M., 2012. How to achieve international action on falsified and substandard medicines. BMJ 7381: 1–6.

6. World Health Organization, 2011. Counterfeit, FAQs. Available at: http://www.who.int/medicines/services/counterfeit/faqs/QandAsUpdateJuly11.pdf. Accessed September 29, 2016.
7. Consortium ACT, 2016. Analysing the Quality and Authenticity of ACT Drugs. Available at: http://www.actconsortium.org/projects/9/analysing-the-quality-and-authenticity-of-act-drugs. Accessed September 29, 2016.

8. Hall Z, Allan EL, van Schalkwyk DA, van Wyk A, Kaur H, 2016. Degradation of artemisinin-based combination therapies under tropical conditions. Am J Trop Med Hyg 94: 993–1001.

9. Game C, Officer CP, 2013. Global ACT Supply Chain Threats Conference. Available at: http://www.theglobalfund.org/en/sourcing/updates/2013-07-09_Innovation_and_Global_ACT_Supply_Chain_Threats_Conference/. Accessed September 27, 2016.

10. Martino R, Gilard V, 2010. Counterfeit drugs: analytical techniques for their identification. Anal Bioanal Chem 398: 77–92.

11. Markert H, Ring J, Campbell N, 2011. A comparison of four commercially available portable Raman spectrometers. National Forensic Science Technology Center 727: 1–8.

12. Hajjou M, Qin Y, Bradby S, Bempong D, Lukulay P, 2013. Assessment of the performance of a handheld Raman device for potential use as a screening tool in evaluating medicines quality. J Pharm Biomed Anal 74: 47–55.

13. Ricci C, Nyadong L, Yang F, Fernandez FM, Brown CD, Newton PN, Kazarian SG, 2008. Assessment of hand-held Raman instrumentation for in situ screening for potentially counterfeit artesunate antimalarial tablets by FT-Raman spectroscopy and direct ionization mass spectrometry. Anal Chim Acta 3: 178–186.

14. Bate R, Tren R, Hess K, Mooney L, Porter K, 2009. Pilot study comparing technologies to test for substandard drugs in field Settings. Afr J Pharm 3: 165–170.

15. Ranieri N, Tabemero P, Green MD, Verbois L, Herrington J, Sampson E, Satzger RD, Phouavong C, Thao K, Newton PN, Witkowski MR, 2014. Evaluation of a new handheld instrument for the detection of counterfeit artesunate by visual fluorescence comparison. Am J Trop Med Hyg 91: 920–924.

16. Tivura M, Asante I, van Wyk A, Gyaase S, Malik N, Mahama E, Hosteller DM, Fernandez FM, Asante KP, Kaur H, Owusu-Agyei, 2016. Quality of artemisinin-based combination therapy for malaria found in Ghanaian markets and public health implications of their use. BMC Pharmacol Toxicol 17: 48.

17. Kauss T, Fawaz F, Guyot M, Lagueny AM, Dos Santos I, Bonini F, Olliaro P, Caminiti A, Millet P, 2010. Fixed artesunate–amodiaquine combined pre-formulation study for the treatment of malaria. Int J Pharm 395: 198–204.

18. Lacaze C, Kauss T, Kiechel JR, Caminiti A, Fawaz F, Terrassin L, Cuart S, Grislain L, Navaratnam V, Ghezzoul B, Gaudin K, White NJ, Olliaro PL, Millet P, 2011. The initial pharmaceutical development of an artesunate/amodiaquine oral formulation for the treatment of malaria: a public-private partnership. Malar J 10: 142.