

Pathways, outputs and impact of NIH-supported bioinformatics and genomics graduate trainees in Africa

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Abstract

Global biomedical and health research is increasingly relying on genomic and computational approaches, largely driven by the increasing volumes of nucleic acid sequencing. Concurrently, epidemiological studies and clinical records are generating enormous amounts of data amenable to disease modeling, machine learning, and artificial intelligence techniques. Bioinformatics and data science expertise is therefore essential for improved population health. Accordingly, in 2012, the US National Institutes of Health (NIH) in partnership with the Wellcome Trust, and with support from the African Society for Human Genetics, initiated the H3Africa (Human Heredity and Health in Africa) consortium. One of its key goals was to build capacity among African scientists to lead research on genetic and environmental contributors to health and disease across the continent. In 2017, the NIH provided funding to support the establishment of four graduate bioinformatics training programs across five African universities. Over seven years, these programs enrolled multiple trainees ($n > 270$), with >110 earning Master's degrees and >20 completing PhDs in Bioinformatics. It is thus timely to evaluate the outcomes and impact of these programs, particularly regarding graduation rates, career trajectories, and the institutions and research domains their alumni are serving. We also assess employment outcomes and the nature of the research they are enabling ($n > 110$ peer-reviewed articles). We additionally include the progress and outputs of the programs' instructors, which were partially enabled by program resources, networks, and trainees. Overall, this review paints valuable insights into the pioneering role of NIH extramural support in shaping Africa's biomedical research landscape.

Keywords post-graduation pathways, bioinformatics, genomics, graduate trainees, Africa

Introduction

Africa and low- and middle-income countries (LMICs) in general bear a disproportionately higher burden of killer infectious diseases such as Malaria, leishmaniasis, tuberculosis (TB), and human immunodeficiency virus in terms of both morbidity and mortality. Furthermore, the continent has a high endemicity of neglected parasitic human and animal diseases like schistosomiasis and trypanosomiasis [1–3]. Moreover, we are seeing an increasing incidence of non-communicable diseases (NCDs) like diabetes and cancer on the continent [4, 5], owing to longer life expectancy, urbanization, and changes in lifestyle and

diet. Acute shortages of healthcare workers, health facilities, and sub-optimal biomedical research exacerbate these problems. Biomedical research and its translation into public health is progressively becoming dependent on computing and computational platforms (bioinformatics, data science, and artificial intelligence), which portend substantial improvements in efficiencies of health processes, accuracy and speed of research.

Bioinformatics is a branch of data science that involves the acquisition, storage, analysis, interpretation and dissemination of biological

and health data, most often molecular sequencing data on macromolecules essential to how life functions like deoxyribonucleic acid (DNA), ribonucleic acid (RNA) and proteins. It sits at the interface between clinical and basic science research, as it draws upon concepts of population science and the knowledge and theory of biological and computational science. Because it is a so-called ‘dry’ science conducted largely within the computational space, it tends to be the cheaper, quicker and more scalable research component. Indeed, the last twenty years have seen the migration of a significant component of biomedical research from the wet laboratory to the ‘dry’ computer [6–8], enabled by the simultaneous technological developments in high-throughput sequencing [9, 10] and increased computational power. The relatively cheaper cost and faster execution of bioinformatics as a digital platform for research in comparison to wet laboratory experimentation is particularly appropriate for resource-limited settings since it is consequently more scalable. Furthermore, the increasing potential and corresponding importance of data science and artificial intelligence in biomedical research [11, 12] underscores the importance of building capacity in these data-intensive sciences within Africa to help leap-frog the continent from its relative lag in basic, clinical and biomedical research [13, 14]. Yet, sufficiently skilled bioinformaticians are still few and far between within LMICs, which constitutes a critical gap for the achievement of the scientific promise of bioinformatics, data science and artificial intelligence in health.

In 2017, the National Institutes of Health (NIH), through its Fogarty International Institute (FIC) and funding from the NIH Common Fund, announced five-year support to start four bioinformatics and genomics degree training programs across Africa. These were, respectively: Nurturing Genomics and Bioinformatics Research Capacity in Africa (BRECA) based at Makerere University in Uganda, the East African Network for Bioinformatics Training (EANBiT) based at the International Centre of Insect Physiology and Ecology (ICIPE) in Kenya but coordinating training in Kenya and Uganda, the West African Sustainable Leadership And Innovation Training In Bioinformatics Research (WASLITBRe) hosted at Covenant University (CU) in Nigeria but additionally working with the Kwame Nkrumah University of Science and Technology (KNUST), Ghana, and University of Sciences, Techniques & Technologies of Bamako (USTTB) in Mali and the West African Center of Excellence for Global Health Bioinformatics Research Training (WACE-B) based at USTTB in Mali and working with other African Francophone Countries. Many of these groups were members of H3ABioNet, a pan-African Bioinformatics Network within the H3Africa initiative, which ran short courses in bioinformatics for 12 years and developed guidelines for starting bioinformatics degree programs early in the project [15, 16]. H3ABioNet also contributed to faculty and mentorship for the new degree programs. Although several other bioinformatics trainings, usually in the form of short-term workshops or courses ranging from a few days to a couple of months have been and continue to be carried out on the continent [17, 18], the programs described here stood out in two important ways. First, it was specifically designed for long-term (at least two years) degree training, recognizing as it did the need for in-depth, sustained instruction and mentorship as prerequisites for the creation of a professional, research-competent bioinformatics workforce. Secondly, it was executed as multiple programs, spanning multiple universities across different countries on the continent. It thus took on an unprecedented level of scale in terms of duration of training and geographical scope. Although the programs were largely autonomous, they all received broad oversight from the NIH. They maintained a level of

coordination through shared materials, biannual consortium meetings and sometimes shared instructors and curriculum discussions.

Justification for long-term bioinformatics training

Bioinformatics scientists are individuals who can apply bioinformatics and data science principles as independent scientists—with the ability to conceive and answer critical biomedical questions using sound scientific principles with relevant bioinformatics techniques [19]. They can be contrasted with bioinformatics engineers or technicians, who can apply bioinformatics techniques to specified problems but may not be well grounded in biological and biomedical principles to conceive and perform independent research [19]. Skills for the latter can often be gained in short-term training of days, weeks or a few months. Bioinformatics scientists, on the other hand, need to be thoroughly grounded in all major biological, bioinformatics and data science domains [20, 21], including Next Generation Sequencing (NGS), sequence alignment and analysis, phylogenetics, bioinformatics programming (python, R and Unix), primary and secondary bioinformatics databases, molecular biology, population genetics, molecular evolution, and disease modeling. That range of knowledge and skills, together with the required depth, demands sufficient time [18, 20, 21] for learners to digest, contextualize, discuss and practice skills and knowledge in stratified sequential phases that often take years. Such learning is best delivered in longer-term in-person settings, as is the case for high schools [18, 22–25], undergraduate [26, 27], graduate degree, and postdoc programs [28]. It requires years of commitment from both the learners and the instructors. It is thus more susceptible to resource constraints like the paucity of long-term expert trainers and/or mentors, sustained computing infrastructure and effective accreditation bureaucracies. Indeed, this is evidenced by only a few such programs in Africa [28] and elsewhere [29]. The goal and support from the Fogarty International Center and NIH Common Fund [30, 31] were thus to initiate and sustain fully accredited long-term bioinformatics education degree training programs in Africa. Be that as it may, some non-degree training was also supported.

Master of Science versus Doctor of Philosophy training rationale

The programs were designed mostly for MSc bioinformatics training to meet the urgent need for those skills in Africa, and to achieve an initial critical mass of bioinformaticians and data scientists in a reasonable turn-around time (2 years). The choice was also informed by the relatively cheaper cost of MSc training relative to PhD training. Furthermore, PhD training requires a master’s degree, so we needed to form a large pool of master’s degrees to eventually feed into the PhD programs. The average PhD:MSc cost ratio of ~3:1 would mean the attainment of at least thrice more trainees [32]. In our cohorts, the ratio of MSc to PhD enrollees was 4:1. Nonetheless, PhD training, which takes much longer and is more expensive, remains essential for engendering more advanced bioinformatics and data science independent research leadership [33], and was thus also supported on some of the training programs, albeit on a smaller scale (fewer numbers).

Post-graduation pathways

These graduate programs have been successfully started and have taken root in East and West African Universities. Their establishment

has included substantial investments by way of careful planning, funding, curriculum development, recruiting, retaining and growing a cohort of instructors, careful recruitment of trainees and extensive efforts in instruction and mentorship [34–36]. The information covering those programs' development, nature and delivery have been well documented elsewhere [37–40]. However, as with all education and scientific initiatives, the ultimate value has to be measured by first the rates of matriculation and graduation from the programs, second, the post-graduation pathways and placements of trainees and ultimately, their impact on the biomedical research and health landscape. Those post-graduation parameters constitute the focus of this paper.

Materials and Methods

The broad training program was initiated under the Human Heredity and Health in Africa (H3Africa) [41–43] Global Health Bioinformatics Research Training Program. The goal was to build genomics research capacity by supporting bioinformatics research training programs at LMIC institutions within Africa. It emphasized partnership with other African or high-income country (HIC)-based collaborators to build sustainable centers of bioinformatics research training relevant to global health research for the African continent.

Specifically, four training programs were initiated and supported, supporting training at five universities spanning five countries in East and West Africa, with the support of a coordinating center in South Africa (Fig. 1). The support included trainee tuition, stipend, international research placements for PhD students, salary support for instructors, curriculum formulation, training materials, connectivity costs and annual consortium meetings. The four host universities were, in turn, supported by three US universities (Baylor College of Medicine, Georgia Institute of Technology and Tulane University) which provided up to one year of research placements and mentorship for PhD students. The training programs were respectively titled: BRECA based at Makerere University in Uganda; the EANBiT based at the ICIPE in Kenya with training at Pwani University; the WASLITBRE based at CU in Nigeria with additional training at the KNUST, Ghana; and the WACE-B based at the USTTB in Mali. All programs were duly accredited by the local accreditation bodies. Training was supported beyond this core network of host institutions, and there were some shared activities between some programs and countries, especially around curriculum design. The EANBiT training program supported the initiation of a bioinformatics MSc program at Pwani University, supported trainees within the Makerere University program, and offered scholarships to several Ugandan students tenable at Pwani University in Kenya; EANBiT and BRECA held joint internship programs and joint curriculum review meetings [37]. WASLITBRE supported trainees at CU in Nigeria, USTTB in Mali and KNUST in Kumasi, Ghana. Makerere University's BRECA trained some students from Eswatini, formally the Kingdom of Eswatini and Kenya. The principal investigators, co-investigators and trainees from all programs met twice a year at the consortium bi-annual meetings facilitated by the coordinating center to share progress, recruitment and training, and jointly evaluate approaches as well as curriculum updates. Together, these accredited training programs constituted the first large-scale, transnationally coordinated and sustained higher-degree level training in bioinformatics, which inevitably required significant resources.

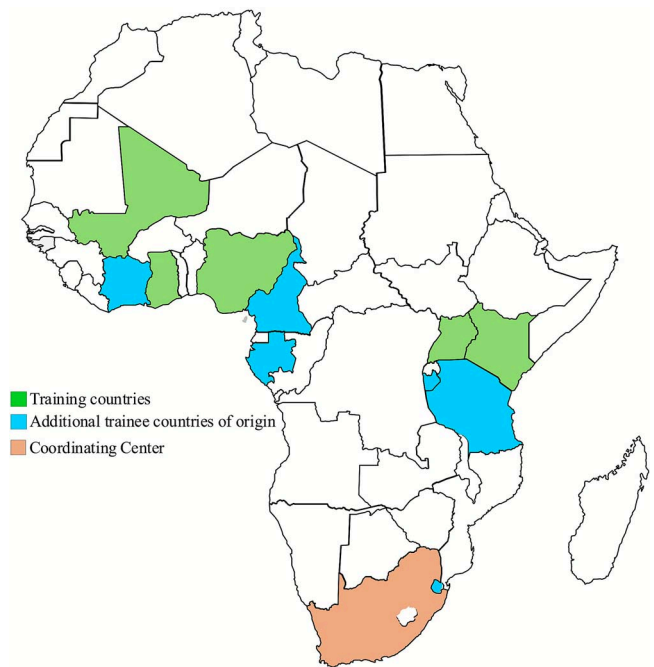


Figure 1 Map of training countries (green), coordinating center (orange), and additional countries of origin for some trainees (blue).

Data gathering

For each of the four training programs, principal investigators and instructors collected four types of data related to the program outputs: (i) they compiled data on the numbers of trainees who had enrolled in both the MSc and/or PhD courses on their programs, (ii) compiled data on numbers of trainees who have graduated from each of those levels, (iii) contacted program graduates and collected data relating to current or previous jobs or placements held by the trainees post-graduation—their roles, institutions and focus of their work by domain, (iv) data on research being performed by various graduates as evidenced by journal publications, some of which were identified using electronic searches like PubMed (grant number or individual name as a key words).

Data analysis

The above-described data were then compiled, curated, collated, categorized, ranked and visualized using R (v4.0.5) programming software to appropriately illustrate the various dimensions of outputs and impact from the training programs. For enrollment and graduation, the numbers of enrolled trainees from each of the five training sites/universities were compiled for each year the program has been running—for both the MSc and PhD trainees. For graduation, a similar compilation was then done for each of the five training sites/universities, starting 2–3 years after matriculation, when the trainees had started graduating (Table 1).

Tracking post-graduation pathways and placements

To establish the postgraduation pathways, the principal investigators and instructors of all four programs respectively conducted a survey that reached out to not only program graduates but also to all persons who had initially enrolled in the programs. Information was collected

Table 1 Number of trainees enrolled and graduated for each training site at the different levels (MSc or PhD) for the different years. Trainees that matriculated from 2018-2021 for MSc and 2018-2020 for PhD are due for graduation and those that matriculated 2022-2024 for MSc and 2021-2024 for PhD were still in training at the time of this assessment.

MSc													
Enrolled									Graduated to date				
Training sites	2018	2019	2020	2021	2022	2023	2024	Total	2021	2022	2023	2024	Total
BRECA-Uganda		28	16	12	13	17	20	106		19	7	12	38
EANBiT-Kenya	8	12	12	12	0	11	7	62	8	7	10	17	42
WACE-B-Mali		3	5	10	10	2		30	10	10	2		22
WASLITBRE-Nigeria		1		2				3			1		1
WASLITBRE-Ghana	0	4	0	5	0	10		19		4		5	9
Total/Year	8	48	33	41	23	40	27	220	18	40	20	34	112

PhD													
Admitted									Graduated to date				
Training sites	2018	2019	2020	2021	2022	2023	2024	Total	2021	2022	2023	2024	Total
BRECA-Uganda		6	1	3	3	4	4	21			2	2	4
EANBiT-Kenya													
WASLITBRE-Nigeria	8		1	2				11	1	2	4	2	9
WACE-B-Mali			5	5	5			15				4	4
WASLITBRE-Ghana	0	4	0	1	0	1		6				3	3
Total/Year	8	10	7	11	8	5	4	53	1	2	6	11	20

about where (institutions or industries) these graduates were now working and brief descriptions of research projects or industry programs they were working on.

Measuring research impact of graduates

Beyond establishing the post-graduate placements of the trainees, we also attempted to measure the extent of their impact, particularly in research informing biomedical and public health progress. This was measured by peer-reviewed publications they had authored or co-authored prior to or after graduation. Information about research domains impacted by their work, their institutional affiliations, diseases investigated and extent of cross-country collaborations was then abstracted from the publications.

Results

Trends of enrollment

The five training sites/universities had different rates of admissions and graduations over the 7 years under assessment. All programs used at least the first year of the grant period (2017–2018) to develop and pursue accreditation of the curricula, recruit instructors and advertise the programs. Consequently, actual instruction started after the 1st or 2nd year, depending on how fast the accreditation process was at the different universities. Hence, graduation did not start until the 4th year of the program. Over the seven years, the cumulative number of trainees admitted to all programs totaled 220 for the master's and 53 for the PhD. The close to 4:1 ratio of master's versus PhD fellows is to be expected, given that the doctoral training is more expensive, takes significantly longer, and has the completion of a master's as one of the requirements for admission, resulting in a substantially smaller pool of qualifying applicants. On average, ~35% of admitted trainees were female, about one in every three trainees. This is a comparatively remarkable percentage, given that in Africa, the rate of females in

science, technology, engineering, and mathematics (STEM) graduate training is significantly lower [44].

Trends of graduation

Ninety (90) of the 220 trainees admitted to the masters program joined in the last three years and are thus still under training. However, of the 130 trainees admitted for the master's program over the first four years (and should therefore have graduated), a total of 101 have graduated across the five training sites. This represents a graduation rate of ~77%. Several factors account for the remaining 23% that have not graduated. Some, especially among the self-supporting trainees, dropped out of the training due to financial constraints (9%). Others are taking longer to complete their research projects due to delayed experiments or data acquisition and are therefore still in the training pipeline (5%). A few, for various personal and family reasons, deferred their studies (9%). Notably, the graduation rates among those with fellowships are significantly higher at ~90%. This sub-group not only has fewer financial roadblocks but is also pre-selected by way of taking applicants with the highest potential for competitive fellowships. Regardless, the 77% overall completion rate, particularly in STEM, compares favorably to graduation rates elsewhere [45], such as 74%–77% in STEM fields at U.S. public research universities [46]. Of the 25 trainees admitted for PhD over the first three years of all the programs and should therefore have graduated, 17 have graduated. This represents a graduation rate of 68%, which compares well with graduation rates elsewhere. This is likely because all PhD trainees had fellowships of some sort and were thus unhindered by financial constraints. Also, doctoral fellows tend to be more committed and focused, a criterion assessed during selection for the scholarships through personal statements and oral interviews. The remaining 32% is almost entirely accounted for because their research projects are taking longer due to experimental validations, data acquisition and data analysis delays. For a few, it is taking longer to get their publications through the peer review process—at least

at GeneNetwork at the University of Tennessee Health Science Center. In Nigeria, a graduate works as an infectious disease modeler at Corona Management Systems and an adjunct researcher at CU. These institutions were previously hard-pressed to do this type of work, and typifies the tracks these graduates are taking. The complete list of placements and current work of graduated trainees can be seen in [Appendix A](#).

Institutions impacted or served by graduates of the programs have ranged from research institutes, public health laboratories and universities to hospitals and ministries of health. Of note is that some graduates have worked across a multiplicity of institutions simultaneously or serially, partly because their skills are still scarce in the ecosystem. Thus, several new linkages and institutional partnerships have been enabled or strengthened ([Fig. 3a](#) and [b](#)).

Beyond the impact the program graduates are having at their respective destinations, the programs themselves have left substantial capacity at their various universities in their wake. MSc and PhD programs in Bioinformatics have been accredited and established. A critical mass of bioinformatics instructors is building up as graduates from the programs pick up instructorships and lecture-ships on the program or instructors from elsewhere are attracted either as full-time or adjunct staff. Currently, staffing levels are: 18 at Makerere University, 9 at Pwani University, 10 at USTTB, 7 at CU, and 12 at KNUST. Significantly, the programs were institutionalized and have continued to be self-sustaining after the expiry of the grant periods, either through fellowships to trainees from other funders or by direct tuition payments from self-sponsoring students. In Makerere University for example, 65% of current enrollees are self-sponsored, while 35% are sponsored by independent grants or institutions. Unsurprisingly, the programs have proved a boon to the instructors themselves, providing a platform for enhancing their pedagogical and research skills, a source for junior colleagues to support their research projects, and a network to establish new collaborations and find research grants. Between them, instructors in these programs have won awards for >24 research projects or training programs ([Appendix B](#)), and several have attained promotions or new appointments, at least in part due to their work, network and support from the programs. This improvement in instructor profiles is an essential dimension of the programs' influence on local research capacity building.

Research contributions

The research outputs and consequent research impact by program graduates and instructors are numerous and diverse, especially as measured by the peer-reviewed research publications they have authored or co-authored. They currently exceed 110 publications, covering four broad domain ([Table 2](#)).

About 71 of those papers are about diseases of importance to Africa, 28 papers relate to bioinformatics methods (software, pipelines and algorithms), 13 are reports on bioinformatics programs—training approaches and frameworks, and six relate to the ethics and policy considerations of the field's application in health. Of note, these research outputs were done in largely collaborative contexts, where program participants were first authors on a majority of the papers (53%), second authors on 37% and last authors on 37% ([Table 2](#)). This represents a substantial contribution to that work and fairly equal collaborative relationships ([Fig. 4](#)).

Research on diseases of importance to Africa

As might be expected, and perhaps appropriately so, a considerably higher volume of the graduates' research has leaned toward diseases of importance to the continent. It ranges from those responsible for the highest morbidity and mortality (Malaria, HIV and TB) to neglected tropical diseases like leishmaniasis. In between are emerging and re-emerging diseases and pandemics like COVID-19, viral hemorrhagic fevers like arboviruses and rhinoviruses; a range of NCDs of growing prevalence like diabetes and high blood pressure (Hbp), as well as NCDs that tend to have a unique prognosis and outcomes among Africans such as cancer, preeclampsia and kidney disease ([Fig. 5](#)). Covered on a lower scale are neural and mental disorders like epilepsy, autism, post-traumatic stress disorders and meningitis ([Fig. 5](#)). An attempt has been made by program alumni to evaluate the extent of genomics resources in Africa, providing a catalog of the current landscape of sequenced and publicly shared pathogens in different countries on the continent [[52](#)].

Malaria

For malaria, research findings have ranged from the establishment of a treatment protocol for the prevention of uncomplicated *P. falciparum* malaria in the demographic where most mortality lies—children between 3 months and 9 years old [[53](#)]; a description of the contribution of travel to malaria transmission [[54](#)]; the identification of factors associated with the occurrence of clinical malaria across different ecological settings in Mali [[55](#)] which could help the development of new strategies for malaria elimination; and the establishment of a multiplicity of infections in asymptomatic malaria patients [[56](#)]. Several studies have been performed in the malaria therapeutics space for potential drug targets [[57–64](#)], including the use of pharmacophore modeling to identify important potential lead compounds that could serve as inhibitors of Pf 5-ALAS [[56](#)] and the use of molecular docking techniques to identify potential inhibitors of PfPPTS [[62](#)]. Other studies have explored Anopheles and Plasmodium metabolic products, pathways and gene usage as well as transmission, development and function of the vector and parasite [[65–69](#)].

Tuberculosis

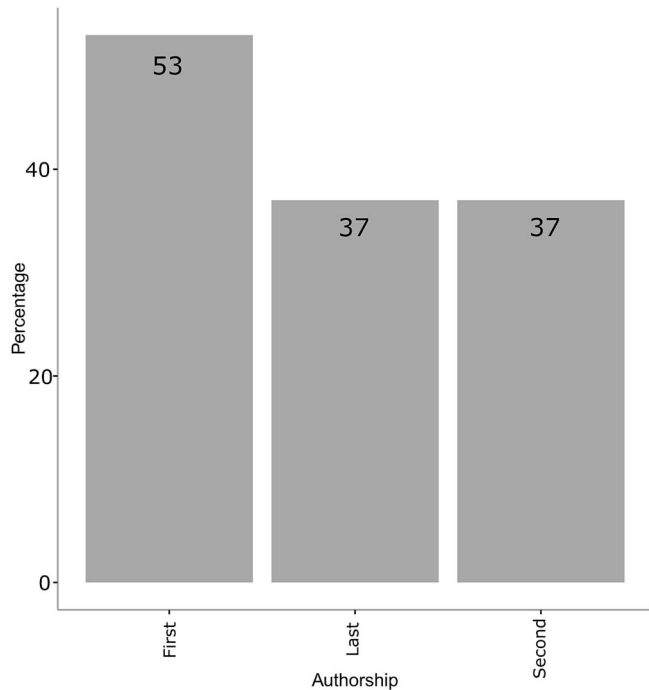
TB research has included a review examining the current progress in studying polymorphisms within immune genes associated with TB susceptibility, focusing on African populations [[70](#)]. It has also investigated the impact of *Mycobacterium tuberculosis* (*M.tb*) infection and *M.tb*-specific IFN γ immune responses on airway microbiome diversity. This was done by performing TB GeneXpert and QuantiFERON-GOLD assays during the follow-up phase of a longitudinal HIV-Lung Microbiome cohort of individuals recruited from two large independent cohorts in rural Uganda [[71](#)]. Closely related work found that airway microbiome signature accurately discriminates *M.tb* infection status [[71, 72](#)]. Further work is now leveraging machine learning to predict antibiotic resistance in four anti-TB drugs (rifampicin, isoniazid, streptomycin, and ethambutol) as part of disease surveillance [[73](#)].

Human immunodeficiency virus

Work in HIV has been varied, ranging from a mapping of the distribution of HIV subtypes in several African countries [[74](#)] to the molecular epidemiology and phylogenetics-based studies of transmission, including the finding that men disproportionately contribute to HIV

Table 2 Broad categories of domains covered by graduates' work.

Domain	Publications	Percentage
Diseases	71	60.2
Bioinformatics methods and tools	28	23.7
Bioinformatics capacity building programs	13	11.0
Ethics and policy	6	5.1

**Figure 4** Percentage of first, second, and last authorships contributed by program participants on collaborative publications ($N = 110$). Participants most frequently appeared as first authors (53%), followed by second (37%) and last authors (37%), indicating strong involvement in both lead and senior research roles.

developing a pipeline that detects mutations in HIV RNA-Seq data [48] and another that performs HIV drug resistance identification, variant evaluation, and surveillance [47]. Another tool integrates NGS, clinical and demographic data for analysis of HIV drug resistance [49]. Further methods-based work applied fine-scale network analysis to estimate transmission network parameters of HIV sequences from key populations, revealing the potential value of such analysis in predicting the progression of the HIV epidemic [50]. Additional research has explored the use of metagenomics as a tool to study non-malarial febrile fevers in HIV-infected children [78].

Emerging and re-emerging infections

Research by program fellows has fittingly investigated some emerging and re-emerging diseases and pandemics on the continent. Even though Africa had comparatively fewer COVID-19 fatalities [79], the pandemic received some research attention from program fellows. For example, there was an assessment of the dynamics of circulating SARS-CoV-2 variants identified during the different COVID-19 waves

in Mali [80]. Similarly, the need for continuous surveillance of SARS-CoV-2 virus to detect emerging variants was illustrated in Uganda [81]. Relatedly, establishing the demographics and seroprevalence of the virus in vaccine-naïve participants was performed in the Democratic Republic of Congo (DRC), Guinea, Liberia, and Mali [82]. There have also been attempts to research potential therapeutic targets for the disease, including an application of immuno-informatics that identified four proteins with advantageous binding to the relevant human HLA-1 and a use of virtual screening of African natural products databases to identify high-affinity inhibitors of SARS-CoV-2 main protease [83]. Beyond the biology of COVID-19 itself, the lockdowns during the pandemic provided an opportunity to confirm that air quality can be used as a proxy for population mobility [84] and to determine the effectiveness of virtual reality (VR) as a tool for training frontline health workers in the context of highly infectious diseases like COVID-19 [85].

Work on other emerging and re-emerging diseases has highlighted the need for systematic molecular studies to understand the transmission of Q fever [86] and the development of a type-specific full genome sequencing approach for obtaining Human Rhino Virus A (HRV) genomic data to characterize infections [87]. For arbovirus infections, a program graduate has contributed to identifying early infection molecular signatures that can differentiate patients who will progress to severe dengue infection. These findings could pave way for evaluating of immunomodulatory therapies for dengue [88] (under review). Additional work has described mosquito-borne viruses in Uganda, focusing on their discovery, morbidity, mode of transmission, animal hosts and clinical manifestations when involved in disease outbreaks. Work on other common viral infections has covered developing and validating a method for detecting and quantifying hepatitis b virus (HBV) DNA by qPCR [89] and, importantly, using NGS to highlight the presence of undiagnosed viruses causing measles-like illnesses (MLI) in Uganda, including vaccine-preventable illnesses. Such findings inform diagnostic assay selection and vaccination strategies [90].

Non-communicable diseases

A range of NCDs of growing prevalence like diabetes and Hbp, as well as NCDs that tend to have a generally distinct prognosis and outcomes among Africans such as cancer, preeclampsia, sickle cell disease and kidney disease have been investigated. For cardio-vascular diseases for example, multivariate meta-analysis has helped identify novel genetic loci associated with cardiometabolic risk factors (CMRFs) in two continental African populations—Ugandans and South Africans [91] and some genetic variants associated with Hbp traits have been identified [92]. Surveys have established the prevalence of type 2 diabetes (T2DM), community awareness of the condition and its attendant risk factors in rural Mali [93], as well as the development of a set of materials for its management [94]. Molecular dynamics simulations have been used to gain insights into the conformational stability of the insulin receptor gene to better understand the physiology of the disease [95]. Its intersection with kidney disease has been investigated through a determination of whether inhibition of kidney function drug targets adversely impacts T2DM [96] as indeed has its intersection with dyslipidemia [97]. Genetic variants associated with lipid traits in African populations have been identified [98]. For kidney disease, which tends to have a unique prognosis among African ancestry populations, binding sites within the APOL1 protein that could be an attractive site for potential inhibitors of APOL1 in the

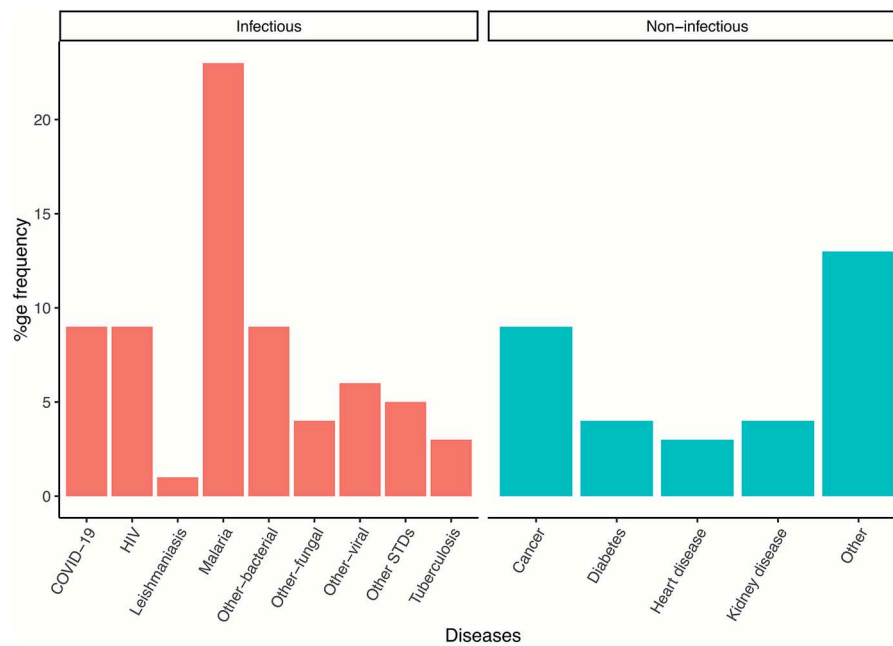


Figure 5 Relative frequency of research on different diseases of public importance on the continent.

management of chronic kidney disease were located [99] as were TrkB agonists for the treatment of CDKL5-deficiency disorders [100]. Uganda's largest patient consented registry was established for sickle cell disease, enrolling patients across four regional referral hospitals. This registry supports studies on clinical phenotypes, outcomes, and interventions like hydroxyurea use and newborn screening showcasing the feasibility of longitudinal disease registries in resource-limited settings [101].

Cancer

As a growing epidemic on the continent [102–104], cancer has received a decent amount of attention from program graduates. That ranges from an assessment showing low knowledge and perceptions of urinary blood cancer among dye workers in Ghana [105] to a review of studies that used radiation therapy as a treatment for breast cancer in Africa, focusing on survival outcomes, adverse effects, radiation therapy techniques, fractionation schedules, and effectiveness of radiation therapy [106]. Cancer studies have similarly explored the different clusters of microbial communities that form different functional groups in gastric cancer [107] and antimicrobial drug resistance, exemplified by the genotypic and phenotypic resistance to third-generation cephalosporin in hematologic cancer patients [108]. Other studies have identified potential biomarkers that can be used for the prognosis of Burkitt's lymphoma [109], first described in Uganda by Dennis Burkitt as a lymphoid malignancy involving mostly the jaw of children from equatorial Africa [110]. Further cancer-related work identified essential genes that could be exploited as potential biomarkers for prognosis and therapeutic targets in hematologic malignancies [111].

Maternal health and mental disorders

The training has additionally supported some nascent work on maternal health, for example the identification of three candidate biomark-

ers, sFlt1, sFlt1/PIGF ratio and PIGF, for the diagnosis of preeclampsia in pregnant women [112] and determination of the prevalence of osteoporosis in African postmenopausal women [113]. Work on mental disorders has covered a review of functional studies and the clinical impact of the pathogenic mutations in the sodium channel genes that contribute to epilepsy [114]. Additional research explored the association of plasma protein levels with major depressive disorder (MDD) in multi-ancestry studies across diverse populations [115]. Similarly, analytical work has been applied to identify genes that significantly contribute to autism and a corresponding classifier has been developed to predict the severity of the disease in the neonate [116]. Further work on mental disorders has explored the interaction of post-traumatic stress and ischemic stroke, showing differences in the genetic susceptibility to the development of post-traumatic stress and its associated risks for ischemic stroke between individuals of African and European ancestries [117].

Antimicrobial drug resistance and therapeutics

As one of the most existential problems facing public health [118], some work has been devoted to AMR and therapeutics. This has included the development of arylidene-based quinazolin-4(3H)-one motifs as potential antimicrobial drug candidates for a range of bacterial and fungal pathogens—*Staphylococcus aureus*, *Candida albicans*, *Aspergillus niger* and *Rhizopus nigricans* [119]. This is important work since *C. albicans* and *Staphylococcus* species are the most common fungal and bacterial agents isolated from bloodstream infections worldwide [120]. For *S. aureus*, further work has involved an evaluation of its antibiotic resistance patterns and virulence factors in isolates from pleuritis infections of hospitalized patients [121]. Disturbingly, AMR in gonorrhea has been characterized in Ugandan men, alongside poor antibiotic stewardship and near-universal resistance of key antigonococcal antibiotics [122]. Appropriately, a standardized gonococcal antimicrobial resistance surveillance program has been established in the country to help support

population-level interventions [123, 124] including the development of a portable magnetofluidic platform for effective testing of sexually transmitted diseases like gonorrhea and their susceptibility to drugs [125]. More recent work has explored bringing Machine Learning and data-driven approaches to bear on AMR surveillance. Specifically, this has included testing/calibrating the effectiveness of models trained on more readily available European data for predicting AMR in Africa [12] and optimizing antibiotic use to combat antimicrobial-resistant infections in Uganda and other LMICS through Centers for Antimicrobial Optimization Network (CAMO-Net) [126]. There have been various efforts to identify potential drugs or drug targets on the therapeutic front. For example, *in silico* methods like molecular docking have been applied to investigate alternate antibacterial agents for Methicillin-resistant *S. aureus* from a library of benzimidazole derivatives [127] and a computer-aided drug design approach to predict potential inhibitors of *Salmonella typhimurium*—ortho acetyl sulphhydrylase synthase (StOASS) [128]. Simultaneously, microwave technology has been deployed to develop coumarin-based compounds that could be used in drug development [129].

Bioinformatics methods, software, pipelines, and algorithms

Work by program graduates has stretched beyond the etiology, pathology and treatment of diseases to include the development of computational methods, approaches, frameworks and techniques for analyzing various biomedical and public health data. For example, a computational model for the prediction of new drug targets in Plasmodium was developed [130], and machine learning has been variously applied to develop predictive models for identifying 208 dependency factors (HDFs) in *Drosophila* involved in disease pathogenesis [131], predicting essential genes in *Drosophila* [132] and identifying several essential genes across eukaryotes [133] and predicting foot-and-mouth disease outbreaks in cattle [134]. Indeed, a full review of the standard procedures and resources available for predicting essential genes in organisms using machine learning has been done [135]. Further methods work has involved the development of a system of ordinary differential equations to model malaria transmission dynamics [136] and a review of clustering methods for big data [137]. Some work was committed to evaluating imputation panels and tools for HLA genes [138], yet other work reviewed methods for studying host-parasite protein-protein interactions in malaria [139]. That review provides a useful knowledge base. In terms of experimental methods, fellows' work revealed novel insights into the potential use of black soldier fly larvae (BSFL) in the bioconversion of various highly lignocellulosic diets to fermentable sugars for subsequent value-added products such as bioethanol [140]. It has also included determining the value of an elastic tissue stain in assessing vascular invasion in Colorectal Cancer (CRC) [141]. Finally, a stable and highly accurate numerical tool for computing river flows in urban areas was built as a first step toward a tool for flood predictions [142].

Bioinformatics training approaches and frameworks

As part of self-introspection, program research explored the organization, content and delivery of bioinformatics and data science training at different levels. Important work provided guidelines on how to apply competency-based bioinformatics education and training and illustrated how the competencies apply to different knowledge areas

for curriculum development [35, 143]. Another effort designed and implemented a model for bioinformatics training and mentorships in resource-limited settings [39], while another described a mentorship program for improving bioinformatics in Kenya [38]. A similar program involved the Nigerian bioinformatics and genomics network and its impact and importance within and outside that country [144]. Those training frameworks and curricula were used to improve bioinformatics training across the continent, for example within EANBIT, which evolved to meet challenges for experiential learning through mini projects to enhance the acquisition of skills and collaborations [37]. Trainees have also had opportunities to network and connect with others on the continent at continent-wide events such as the 3rd ISCB Africa Student Council Symposium [145]. Besides mentorship programs, hackathons have also been used as tools to teach bioinformatics while solving an actual bioinformatics problem simultaneously [146]. For example, one of the hackathons built two data portals for microbiome and genomics in African populations [146].

Furthermore, having appreciated the need for advanced training in data science and bioinformatics on the continent, there have been efforts to improve the available bioinformatics and data science infrastructure, mainly through public-private partnerships as in the case of the African Centers of Excellence in Bioinformatics and Data-Intensive Sciences (ACE) [147, 148], based in Uganda and Mali, which were integral parts of the training programs. This ACE project aims to enhance computational infrastructure, develop advanced bioinformatics and data science skills among local researchers and students, and provide innovative emerging technologies for infectious diseases research [147]. One such innovative technology has been VR, which has been used to upgrade postgraduate surgical training in resource-limited settings [149] and to train frontline health workers, especially in the context of highly infectious diseases like COVID-19 [85]. The feasibility of other emerging digital solutions including AR, in meeting the challenges of delivering rapid and equitable access to emergency care training at scale has been explored. This has produced specific policy and practice recommendations for health professional education and training to enhance emergency care [150].

Ethics and policy considerations

Research outputs have also covered and considered the important aspect of ethics and policies that accompany the application of such tools as bioinformatics, data science and artificial intelligence to health in general and biomedical research in particular, especially in resource-limited settings. There have been discussions of innovations in the ethical governance of AI, including highlighting areas in urgent need of attention internationally [151]. There has been exploration and advice for data standards, practices and philosophies of AI with their specific interplays in the role of shaping the future of society to ensure that AI systems contribute to a better future for all [152]. Notably, individuals from the training programs have participated in formulating WHO guidelines for use of AI in health [153]. They have further participated in international efforts exploring the ethics of using digital technologies for health promotion, particularly among the youth [154]. In the policy space, there has been work showing the need for African governments to invest in public plant genomics research and applications, supporting bioinformatics platforms and training programs, and stimulating university and industry partnerships in agriculture and medicine [155]. Separate

work has shown the need for governments to invest more in RNAseq-based research in Africa [156]. Equally, complementary efforts have described the current state, barriers and opportunities for translational research and showed the need to integrate it into the healthcare system [157].

Discussion

This work assesses the pathways, outputs and impact of NIH-supported degree-level graduate bioinformatics training programs on the African continent—the first ones of their scale in numbers and geographical spread. It does not attempt to describe the structure, implementation, execution and delivery of those programs since those have been variously documented elsewhere in papers that describe individual programs [37–40]. Here, we assess their net outputs and impact in aggregate, covering all four funded projects established at five universities in five countries. Together, they have enrolled dozens of trainees, several of whom have already graduated, several are still in the training pipeline and many more continue to matriculate into the training programs. The graduated trainees have since spread out into multiple career paths and several institutions, where they have made multiple impactful contributions, especially through research, teaching and science leadership. Those contributions have resulted into various publications of importance to African biomedical research and public health, as well as new research and training programs. It is noteworthy that most of that research and resultant publications have been in collaboration with scientists at other institutions on the continent, and those from other continents, particularly North America and Europe. Strengthening collaborative science has thus been one of the major incidental outputs of these training programs, as has been the formation of a viable community of bioinformatics and genomics professionals able to specialize in different research endeavors while simultaneously leveraging the synergies of cross-continental research and training.

The amount and extent of outputs are testimony to the value of sustained long-term graduate training. Unsurprisingly, such training requires substantial initial funding, an availability of a threshold of initial trainers, elaborate curriculum designs, reliable accreditation bureaucracies, quality assurance systems and the existence of a reasonable qualifying pool of potential trainees. They also benefit tremendously from sustained support by hosting institutions and the vision and enthusiasm of faculty. Indeed, the NIH funded the initiation of the training programs, i.e., curriculum design, staff support, students' fellowships (tuition and stipend), teaching materials, travel and coordination. This amounted to ~\$5.2m [158] over a period of five years across the five training sites, representing ~75% of the total cost of the programs. Supplementary support came from host universities, which provided lecture rooms, utilities, computing platforms like HPCs, and part of the staff effort, all of which cost above the allowable NIH indirect costs. Additional funding to the universities came from self-sponsored students who contributed subsidized tuition fees. These local contributions bode well for the continued sustainability of the programs. Now that the programs are in place, they can be leveraged to support other forms of training. For example, program graduates and instructors have supported other training modes, such as the short-term Introduction to Bioinformatics Training (IBT) organized by H3ABioNet [15, 36] and Africa CDC's sequencing capacity building for public health laboratories across the continent [159]. Those have, in turn, also served as a preparation platform for aspiring candidates

for long-term graduate training, which illustrates the cyclical relationships and synergies such programs have with other forms of training as well as with research and public health endeavors. Obviously, for sustained relevance and continued quality outputs, the programs must be continuously updated to co-opt new growth areas for biomedical research such as artificial intelligence, machine learning and other mostly unmet needs like generative AI and image analysis. Indeed, in subsequent funding, the NIH has supported either new programs for those areas or their integration into existing ones under the Data Science for Health Discovery and Innovation in Africa (DS-I Africa) Initiative. Perhaps one key unexpected dividend of these programs has been the high participation/integration of female trainees, who have accounted for ~33% of participants, above normal levels in STEM graduate training [160]. Indeed, recent global estimates indicate that female participation rates in many African graduate STEM programs typically range between 20%–30% [161, 162], comparable to ~28%–38% of STEM graduate students in the US [160]. Against this backdrop, our observed female participation exceeds typical regional benchmarks and slightly surpasses participation levels in many high-income countries, suggesting meaningful progress toward gender equity in advanced STEM training.

To the extent that outputs often correspond to inputs, the initial substantial level of investment by the NIH for a sustained period of 5 years unsurprisingly generated a significant impact. That initial spark was essential for the progress of bioinformatics and genomics on the continent. To the best of our knowledge, none of the prior short-term trainings had yielded such a level of outputs and impact. Furthermore, the programs have been fortunate to leverage the parallel growth of requisite computational infrastructure—software and hardware [16, 34, 147, 159]—essential for their progression and sustainability [163]. Whereas outputs and impact have been substantial, there is still room for improvement, particularly by optimizing quality internships and placements, which link trainees to ongoing and functional research and development programs during training.

Conclusion

This program assessment has not been without limitations. Despite our best efforts to track graduates, tracking every last one and their outputs has proven impossible. As a result, not all outputs and pathways have been captured, but the overwhelming majority are represented here. At its best, this paper outlines a snapshot in time. This is so, first because there have only been a few cohorts of program graduates at this point and secondly because they are still in the very early stages of their careers. Their outputs and impact will increase in quantity and quality as a function of both time and experience. Additionally, several cohorts are still in the training pipeline and others will matriculate to the programs in succeeding years. Consequently, a future reassessment can be expected to yield more outputs and even more significant impact. The present work offers a pilot approach and framework that can, in the future, be improved or expanded for such an exercise in training program assessment. However, the many outputs outlined in this paper should be better seen as an indication of what a difference appropriately resourced and planned training programs at the correct scale can make and not a suggestion that the training need has been fully met. Five training sites/universities in five countries remain quite few compared to the need, both present and anticipated, as genomics, bioinformatics and data science increasingly take on a bigger share of research, development and clinical practice [6].

Key Points

- The National Institutes of Health (NIH)-supported Human Heredity and Health in Africa graduate bioinformatics programs, launched in 2017 across five African universities, have trained >270 students, producing >110 MSc and 20+ PhD graduates, significantly boosting Africa's bioinformatics and genomics capacity.
- Graduates have transitioned into impactful roles in academia, research institutes, public health laboratories, and industry, contributing to disease surveillance, diagnostics, and cutting-edge biomedical research across Africa and beyond.
- The programs have generated >110 peer-reviewed publications, with research spanning infectious diseases, non-communicable diseases, bioinformatics tool development, ethics, and policy, addressing critical health priorities for the continent.
- A sustainable pipeline of bioinformatics instructors and mentors has been established, with many graduates now teaching and leading research, ensuring program continuity beyond initial NIH funding.
- These initiatives have fostered strong intra-African and global research collaborations, strengthened institutional capacity, and demonstrated the long-term value of sustained, degree-level training for scientific leadership in Africa.

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Authors contributions

Daudi Jjingo (Led the conceptualization of the study, Performed formal analysis, Managed funding acquisition, Led the project supervision, Drafted the original version of the manuscript), Andrew Walakira

(Performed formal analysis, Produced the visualizations in the paper, Drafted the original version of the manuscript), Suhaila Hashim (Carried out data curation), Cisse Cheickna (Carried out data curation), Ronald Galiwango (Carried out data curation, Performed formal analysis, Drafted the original version of the manuscript), Caleb Kibe (Carried out data curation), Florence N. Kivunike (Carried out data curation), Gerald Mboowa (Carried out data curation, Managed funding acquisition, Led the project supervision), Fredrick Elishama Kakembo (Performed formal analysis, Carried out the investigation, Produced the visualizations in the paper, Drafted the original version of the manuscript), Babajide Ayodele (Carried out the investigation), Jean-Baka Domelevo Entfellner (Carried out the investigation), Santie de Villiers (Carried out the investigation), Karen Wambui (Carried out the investigation, Oversaw the project administration), Segun Fatumo (Carried out the investigation), Tinashe Chikowore (Carried out the investigation), John Mukisa (Carried out the investigation), Alfred Ssekagiri (Carried out the investigation), Nicholas Bbosa (Carried out the investigation), Julius Mulindwa (Carried out the investigation), Samuel Kyobe (Carried out the investigation), Mike Nsubuga (Carried out the investigation), Grace Kebirungi (Carried out the investigation), Eric Katagiryra (Carried out the investigation), Savannah Mwesigwa (Carried out the investigation), Ibra Lujumba (Carried out the investigation), Rogers Kamulegeya (Carried out the investigation), Samuel Kirimunda (Carried out the investigation), Stephen Kanyerezi (Carried out the investigation), Shahiid Kiyaga (Carried out the investigation), Ivan Sserwadda (Carried out the investigation), Davis Kiberu (Carried out the investigation), Bernard S. Bagaya (Carried out the investigation), Julius Okwir (Carried out the investigation), Patricia Nabisubi (Carried out the investigation), Grace Nabakooza (Carried out the investigation), Mugume Twinamatsiko Atwine (Carried out the investigation), Ricard Sserunjogi (Carried out the investigation), Rolanda Julius (Carried out the investigation), Mariam Quiñones (Carried out the investigation), Meghan McCarthy (Carried out the investigation), Phillip Cruz (Carried out the investigation), Karlynn Noble (Oversaw the project administration), Christopher J. Whalen (Provided the resources), Darrell Hurt (Provided the resources), Maria Y. Giovanni (Provided the resources), Michael Tartakovsky (Provided the resources), Deogratius Ssemwanga (Provided the resources), John M. Kitayimbwa (Provided the resources), Steven J. Reynolds (Provided the resources), Christopher C. Whalen (Provided the resources), Andrew Kambugu (Provided the resources), Neil A. Hanchard (Provided the resources), Li Jian (Led the project supervision), Peter Amoako-Yirenkyi (Led the project supervision), Graeme Mardon (Led the project supervision), I. King Jordan (Led the project supervision), Samson Pandam Salifu (Led the project supervision), Mamadou Wele (Led the project supervision), Ezekiel Adebisi (Led the project supervision), Jeffrey G. Shaffer (Led the project supervision), Seydou Doumbia (Led the project supervision), David Patrick Kateete (Led the project supervision), Michelle Skelton (Led the project supervision), Nicola Mulder (Led the project supervision), Jonathan K. Kayondo (Led the project supervision), Daniel Masiga (Led the project supervision), and H3A (Led the project supervision)

Supplementary material

Supplementary material is available at *Briefings in Bioinformatics* online.

Data availability

All data supporting the findings of this study are provided within the manuscript and its supplementary materials.

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