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REVIEW

Substance use and substance use disorders in Africa: An epidemiological approach to the review of existing literature

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Abstract

The relationship between man and substances that have abuse potentials, and whose use has been associated with the development or progression of substance use disorders has continued to evolve in terms of geography, economic implications, and time. History shows that local plants with psychoactive constituents can get exported worldwide through global travel, commerce, or even conquest. Time and globalization also change people's relationship with substances of abuse; hence, an area that was initially alien to certain substances might evolve to becoming a trafficking hub, and then a destination. A case in point is Africa where a rapidly increasing prevalence of substance use/abuse and substance use disorder among adolescents and young adults is putting enormous strain on the economy, healthcare system, and society at large. However, there appears to be a paucity of scientific literature and data on the epidemiology, risk assessment, and contributing factors to substance use and the development of substance use disorders across Africa. In this narrative review, we examine extant literature (PubMed, Google scholar, Medline) for information on the prevalence, trends, and



influencers of substance use and the development of substance use disorders. This is with a view of understanding the determinants of substance use and factors that influence the development of substance use disorders in the region, and how this information can be channeled towards developing a comprehensive intervention and treatment program.

Key Words: Addiction; Cannabis; Catha edulis; Datura metal; Drug dependence; Novel psychoactive substances

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Core Tip: Substance use for medicinal and recreational purposes dates back centuries; however, in recent times, substance use is increasingly becoming a global public health crisis. In Africa, there is a consensus that substance use is emerging as a public health crise, but there appears to be a paucity of data on the epidemiology, risk assessment, and contributing factors to substance use and the development of substance use disorders across Africa. Here, we examined the extant literature for information on the prevalence, trends, and influencers of substance use and substance use disorders as it relates to Africa.

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INTRODUCTION

Substance use and substance use disorders are increasingly becoming a global public health crisis, largely due to their increasing prevalence, worsening disability-adjusted life years, and high socioeconomic burden[1]. According to the World Drug Report, 2021[2], approximately 275 million people used drugs worldwide in the preceding year, with another 36 million persons diagnosed with substance use disorders globally[2]. In 2019 alone, substance use disorders were linked to about 18 million years of healthy life lost. Also, about 180000 deaths were directly linked to substance use disorders, while another half million deaths were attributed to illicit drug use[3].

Substance use is generally defined as a patterned use of any substance (including alcohol and/or psychoactive drugs) in quantities (or through methods) that are harmful to the user or others[4]. Substance use is often associated with varying degrees of intoxication, which is associated with alteration of judgment, attention, and perception. The use of alcohol, illicit drugs, and illegal use of prescription medications has been associated with negative impact on the individual's health and productivity, as well as a high socioeconomic burden on the family and society [5,6]. Globally, there is a rapidly rising prevalence of substance use and substance use disorders, with an associated increase in the morbidity and mortality. Also, in Africa, the use of illicit substances such as cannabis (the most widely used substance in Africa, with a prevalence of between 5.2% and 13.5% in West and Central Africa), amphetamine-type stimulants, and benzodiazepines is increasing rapidly[7]. Again, in the last decade, Africa has begun to be recognized as a consumer and a destination for illicit drugs, compared to being previously regarded as mainly a transit zone for these drugs (serving as a link between Latin America and Europe)[8,9]. This reversal of the illicit drug trend is believed to be a contributing factor to the rapid development of substance use epidemic, particularly in the urban centers of Africa.

There is a consensus that substance use (particularly among adolescents and young adults) in Africa is emerging as a public health crises; however, there appears to be a paucity of scientific literature and data on the epidemiology, risk assessment, and contributing factors to substance use and the development of substance use disorders across Africa. Here, we reviewed the extant literature for information on the prevalence, trends, and influencers of substance use and the development of substance use disorders. This is with a view of understanding the determinants of substance use and the factors that influence the development of substance use disorders in the region, and how this information can be channeled towards developing a comprehensive interventions and treatment program.

History of substance use and substance use disorders

Substance use for medicinal, religious, and recreational purposes dates back centuries. The earliest mentions of the use of alcoholic or fermented beverages in Chinese writing dates far back as the 7th



millennia B.C.E[10], although there is also evidence from Sumerian writing (2100 B.C.E) of the use of opium from the poppy plant[11]. While the earliest use had been linked mainly to medicinal and religious purposes [10-12], there are also documentations of their use for recreational purposes [12]. However, since ancient times, humans have recognized the health problems that may be associated with excess alcohol consumption[12-14].

In 2019, the United Nations Office on Drugs and Crime reported an estimated 35 million people having a substance use disorder necessitating treatment^[15]. Surveys and results of prospective studies examining patterns of drug use among the general population has revealed substance use peaks between 18 years and 25 years of age[15,16] with drug use among young people exceeding that of older people[15,17].

History of substance use in Africa: From transit nations to major illicit drug destinations

The African continent has a long history of drug cultivation, production, trade and consumption; and there are also indigenous plants and herbs with psychoactive effects such as cannabis resin (known as hashish in North Africa), Catha edulis (known as Kath in East Africa), and cannabis (known as dagga in Southern Africa) that have also been used traditionally for centuries[18-21]. In the last few years, rapidly growing large-scale trade and recreational use of opiates, synthetic psychoactive stimulants, and prescription drugs are emerging threats in the African continent^[21].

The relationship between North Africa and cannabis has existed for centuries, predating the arrival of the Spanish and French in the 19th century. Also, during the colonial era, cannabis was cultivated in small quantities across the northern Rif mountains in Morocco, and throughout the northern parts of Tunisia and Algeria^[22]. While production was mainly to meet local demands (with some smuggling and exportation to Europe), the era following independence of the different countries in the region saw regulations and laws being enacted and enforced to control the production, sale, and use of cannabis. However, across the four countries (Morocco, Algeria, Tunisia and Libya) that make up the Maghreb (an area also known as northwest Africa), the drug trade has not only continued to grow, it is also evolving. In the last few decades, a region known mainly for the production of cannabis destined for other markets (particularly the European market) has increasingly become an important route for the trafficking of cocaine and different psychotropic pills. Since the beginning of the 21st century, trafficking routes for cocaine, cannabis resin, and psychotropic pills that existed between South America, Africa, and Europe shifted to transect the Maghreb region[22]. This change has been partly attributed to an increase in the demand for drugs in the region, and the perturbations of other transit zones such as the Sahel region, which has become unstable. Most important is the geographic location of the region, being a link between Africa, Middle East and Europe. While drug transit routes through North Africa is increasing, of more importance is the increasing rate of consumption of these drugs in the region. The use of psychotropic drugs, which are very addictive, is nearing epidemic proportions in the region; also, other substances being consumed include cannabis, cocaine and opioids[22].

Before West Africa began to be considered a transit zone for drugs, it was also a producer of cannabis products (although not on the scale of the North African countries), which were shipped to Europe and the United States. Although, at the same time, marijuana was being imported into Nigeria from South Africa and a region now known as the Democratic Republic of Congo^[23]. The smuggling of heroin through West Africa was first documented in 1952[23]; however, West Africa's rise as a major drug smuggling hub began sometime in the 1960s, coinciding with a period of increased demand for illegal drugs, including marijuana which was grown and exported from Nigeria in large quantities to Europe. Despite attempts by governments in these countries to stem the tide of the marijuana export, marijuana trade continued illegally for several years until the demand for newer psychoactive substances such as cocaine and heroin overtook the demand for marijuana^[24]. By the 990th year of the 2nd millennium and the 90th year of the 20th century, West Africa had become a major transit and repackaging center for substances such as cocaine and heroin through a transnational trade route that originated from South America and Asia, to Europe. While drug trafficking through Africa was not new, an intense clampdown on the South-North American trade routes by the United States anti-narcotics strategies and the increase in demand for drugs across Europe saw to the rapid expansion of the West African trade routes in the early 21st century [25,26]. The geography of the West African states (made up of large areas of uninhabited islands and archipelagoes found in countries like Guinea Bissau) eased transit and made detection difficult^[27]. Also, the vulnerable political environment with the presence of civil wars/insurgencies created fertile grounds for the development of criminal networks in the West African sub region [27].

Previously, when compared to West Africa, the drug trafficking routes through the Eastern belt of Africa were less robust; however, in the last few decades of this century, the trend is an increase in the trafficking of and the variety of trafficked drugs through the East African states of Kenya, Uganda, and Tanzania. Specifically, the trafficking of heroin and cocaine through these countries has grown considerably. Trafficking routes begin in Afghanistan where heroin is produced, through Pakistan and then East Africa enroute over Europe. The cocaine transnational trafficking routes began to go through East Africa in a bid to bypass the West African routes that were increasingly being watched by anti-drug trafficking authorities [28]. Also observed was that the increased consumption of these substances coincided with an increase in trafficking and affordability of the drugs.



In South Africa, a country in the southern region of Africa, the trafficking of drugs has increased. There are reports that since the period prior to and following the transition to democracy, there has been an escalation of drug trafficking. Trafficking in these parts has increased as a result of the easing of the strict control of land, air and sea borders, and an increase in international trade that occurred following the reintegration of South Africa into the committee of nations following the end of apartheid. Also, the effective policing of traditional smuggling routes prompting the search for other shipping routes also accounts for the increased trafficking of drugs through South Africa[29]. The increased trafficking is also worsening the substance use problems as a proportion of the drugs trafficked end up on the local market. There have also been reports that drugs such as methaqualone are also produced in clandestine laboratories in the region [30]. Overall, the level of affluence in the region makes it an attractive 'emerging market' for illicit drugs[30-32]. South Africa also has a history of drug use that dates as far back as the 15th century. Cannabis, which is known as dagga in South Africa, has been consumed traditionally for centuries. The cannabis plant was brought to southern Africa by Saheli merchants from eastern Africa and some members of the bantu tribe of central and southern Africa where it has been cultivated since the 15th century. Around the 16th and 18th century, the consumption of cannabis increased significantly[33]. Although initially popular only among the African population, over time, its use extended to the white population of South Africa[30].

Overall, while the current substance use epidemic in the African continent could be linked to the global trend in substance use, the transformation of African nations from mainly transit points in the international drug network to consumer countries would seem inevitable[34]. Also, the rapid socioeconomic changes that have occurred across the different countries that make up the African continent could have facilitated this shift in what can be assumed to be the "normal trend".

EPIDEMIOLOGY OF SUBSTANCE USE ACROSS THE AFRICAN CONTINENT

Across Africa, reports spanning the last two decades show that substance use especially among adolescents and young adults is increasing at alarming rates [19,35-41]. The World Health Organization and the United Nations Office on Drugs and Crime reported, an exponential increase in the per capita consumption of alcohol as well as the cultivation, trade, and consumption of cannabis in most of the countries in Africa, with suggestions that this could inevitably have adverse socioeconomic and public health implications [42-44]. At the time, about 10 countries in Africa were listed among the 22 countries with the highest increases in the use of alcohol and other psychoactive substances including cannabis, tobacco, cocaine, and heroin [45]. In 2013, the United Nations Office on Drugs and Crime World Drug Report estimated that across the African continent, more than 28 million people had a current history of substance use. Cannabis was also reported to be the most commonly used drug on the continent, with the prevalence estimated to be 7.5%, which was almost twice the global average. The use of opioids was also reportedly on the rise[46].

While it has been recognized that Africa is beginning to battle a drug use epidemic, with an estimated 37000 people in Africa dying annually from substance use-associated complications[47-49]; available data for Africa are still either weak or nonexistent. To date, in many African countries, there is still a paucity of national data regarding the epidemiology and patterns of substance use across populations, with available data largely limited to small prospective population studies and retrospective hospitalbased studies.

In West Africa, the paucity of data regarding the prevalence of drug use undermines our ability to adequately understand the full extent of the substance use problem, and how it is creating a public health problem that further threatens the already fragile health system that currently exists. It also creates a false sense of safety, because it fosters the erroneous belief that substance use is under control. However, in the last few years, this trend is becoming more difficult to ignore, because there is now increasing evidence from the increase in crime/criminal behaviors and an increasing need for medical attention that arises from the development of substance use disorders or complications of risky behaviors that are consequences of drug use. In the last decade, in West African countries like Ghana, incident reports from health professionals, lawyers, and law enforcement officers are beginning to show dramatic increases in the domestic consumption of illicit drugs. However, these reports do not adequately portray the scale of substance use problem; because there is a dearth of national figures that can accurately quantify the prevalence of drug use in Ghana or most other West African countries. All of these result in a huge dependence on small-scale cross-sectional studies (Table 1). A 2008 populationbased study conducted among school-going adolescents, reported that the prevalence of any substance use in the preceding 1 mo was 3.6% [50]. The results of an earlier study that interviewed a sample of 894 high school students with a mean age of 17.4 years, reported that the lifetime alcohol use in these cohort was 25.1%; with cigarette use and lifetime marijuana use being 7.5% and 2.6%, respectively. Also, current alcohol use was reported to be 46.2%; current cigarette and marijuana use was 44.6% and 58.3%, respectively^[51]. The result of a 2014 cross-sectional survey of a sample of 227 street children and youths revealed that the current prevalence of alcohol and marijuana use was 12% and 16.2%, respectively. Sex differences in substance use was also reported with more females using alcohol, marijuana, and



Region	Study type	Study group	Result	Ref.
Ghana	Population-based study	School-going adolescents	3.6% prevalence of substance use in the preceding 1 mo	[5 0]
Ghana	Cross-sectional study	894 high school students with a mean age of 17.4 yr	Lifetime alcohol use was 25.1%; with cigarette use and lifetime marijuana use being 7.5% and 2.6% respectively. Current alcohol use was 46.2%; current cigarette and marijuana use was 44.6%; and 58.3%, respectively	[51]
Ghana	Cross-sectional survey	227 street children and youths	Current prevalence of alcohol and marijuana use was 12% and 16.2%, respectively	[<mark>52</mark>]
Nigeria	Cross-sectional study Northwestern Nigeria	280 secondary school students	56% of them had a history of substance use (kolanut, cigarettes, and marijuana)	[<mark>53</mark>]
Nigeria	Cross-sectional study Southwestern Nigeria	249 secondary school students	Prevalence of alcohol and substance use was 21.7% and 26.3%, respectively, tramadol being the substance of choice	[54]
Nigeria	National drug survey	Population-based	Approximately 14.3 million people (accounting for 14.4% of the population aged between 15-64 yr) had a history of current and continuing substance drug use, with close to 3 million having at least a form of drug use disorder	
Ethiopia	Demographic and health survey	Population-based	4% of youths and 6.3% of individuals in age groups of 25-29 yr smoked cigarettes, while 53% of men and 45% of women consumed alcohol	[59]
Ethiopia	Analysis of data extracted from the 2016 Ethiopia Demographic and Health Survey	12688 male cohorts	62.5% (7931 males) had a current history of substance use (alcohol, Kath, or tobacco). Inhabitants of the Amhara, Tigray and Oromia regions had a current substance use prevalence of 18.5%, 14.2% and 12.8%, respectively	[<mark>60</mark>]
Ethiopia	Cross-sectional study Northeastern Ethiopia	730 university students in	Lifetime prevalence of alcohol consumption, Kath chewing, and cigarette smoking was 33.1%, 13% and 7.9%, respectively, and current prevalence was 27.9%, 10.4% and 6.4%	[<mark>61</mark>]
Ethiopia	Cross-sectional study	794 university students	73.7% had a history of substance use with the lifetime prevalence of illicit drugs being 23.3%	[<mark>63</mark>]
Egypt	Hospital-based study (single-center experience)	First episode drug- induced psychosis patients	Substance abuse rates are as high as 10%-20% the global average with cannabis and tramadol being the most abused substance	[65]
Tunisia	Cross-sectional study	298 persons with a history of drug use	Cannabis was the most widely consumed illicit drug, followed by benzodiazepines, buprenorphine, cocaine, and ecstasy	[<mark>68</mark>]
Tunisia	Mediterranean school survey project	Secondary school students	Tobacco, alcohol, and cannabis were the substances most frequently used	[66, 67]
Tunisia	Epidemiologic/toxicological invest- igation Northern Tunisia	11170 suspected drug users	A preponderance of males (97.4%), with a median age of 29 ± 7.91 yr. 91.3 % were single	[<mark>69</mark>]
South Africa	School-based survey	Secondary school students	13% of the students (aged 19 yr and below) had an history of cannabis use, although current use was 9%. 12% had a current use of heroin, 11% used inhalants and 6% consumed mandrax	[73]
South Africa	National household survey	Population-based	Past 3 mo prevalence for cannabis among 15-19-years-old was 3%	[74]

smoking cigarettes compared to males[52]. In Nigeria, reports from small-scale studies have demonstrated a high prevalence of substance use among adolescents and young adults. A 2009 study that examined the prevalence of substance use among 280 students at a senior secondary school in a town in Northwest Nigeria, revealed that about 56% of them had a history of substance use, with the most common being kolanut, cigarettes, and marijuana[53]. Idowu et al[54] also examined the prevalence of substance use among 249 students (mean age = 16.3 ± 2 standard deviations) of secondary schools in a metropolis in south western Nigeria and reported that the prevalence of alcohol and substance use was 21.7% and 26.3%, respectively, with tramadol being the substance of choice[54]. The magnitude of the effect was best conveyed by the results of the 2018 National Drug Use Survey which revealed that approximately 14.3 million people (accounting for 14.4% of the population aged between 15 years and 64 years) had a history of current and continuing substance drug use, with close to 3 million having at least a form of drug use disorder [48,55]. A difference was also observed in the prevalence of drug use between the Northern and Southern geopolitical zones, with a higher prevalence in the regions in the south (13.8%-22.4% of the population) compared to those in the northern geopolitical zone (10%-14.9% of the population). In Nigeria, cannabis was the most commonly used drug, which was followed by opioids (non-prescription or in cough syrup)[48,56]. The survey also highlighted



a rise in the current use of psychoactive substances (including cannabis), the non-medical use of prescription drugs such tramadol, codeine, morphine or cough syrups that contain codeine or dextromethorphan[55]. Also observed was an overall high incidence of drug use (excluding alcohol) among males compared to females (10.8 million males vs 3.4 million females), although the sex difference in the non-medical use of prescription opioids, cough syrups, and sedatives was not as significant (6% among men compared to 3.3 among women). The survey also reported a higher incidence of drug use among young adults (24-39) compared to those aged 24 and below [55].

In East Africa, there is also a dearth of national statistical data on the prevalence of substance use in a number of the countries, with researchers and policy makers needing to rely on information from studies involving subsets of the populations. In Ethiopia, alcohol, Kath and tobacco are the most popular substances that are consumed [57,58]. A 2012 Ethiopian demographic and health survey reported that 4% of youths and 6.3% of individuals in age groups of 25-29 years smoked cigarettes, whereas 53% of men and 45% of women consumed alcohol[59]. Also, the results of a study by Girma et al[60] that analyzed data extracted from the 2016 Ethiopia Demographic and Health Survey revealed that of the 12688 male cohorts of the Ethiopian Demographic and Health Survey, at least 62.5% (7931 males) had a current history of substance use (alcohol, Kath, or tobacco) as at the time of the survey. Inhabitants of the Amhara, Tigray, and Oromia regions have a current substance use prevalence of 18.5%, 14.2%, and 12.8%, respectively. Alcohol (53.1%) is reportedly the most commonly consumed substance, followed closely by Kath, which has a prevalence of 25.9% [60]. Reports of small crosssectional studies have also corroborated the high prevalence of alcohol, Kath, and cigarette smoking among Ethiopian youths[61]. Adere et al[61] examined a cohort of 730 university students in Northeastern Ethiopia and reported that the lifetime prevalence of alcohol consumption, Kath chewing, and cigarette smoking was 33.1%, 13%, and 7.9%, respectively, whereas the current prevalence of these substances is 27.9%, 10.4%, and 6.4%, respectively[61]. The prevalence observed in this study was similar to that observed in an earlier study carried out among the students at a University in a town in North Ethiopia[62]. While earlier studies among university students did not report evidence to suggest the use of illicit drugs, the results of a 2021 cross-sectional study among 794 students of Addis Ababa University, showed that 73.7% of the study participants had a history of substance use with the use of illicit drugs having a lifetime prevalence of 23.3% [63]. However, similar to other studies, alcohol, Kath, and cigarettes were still the most commonly abused substances[63].

In North Africa, data and information on substance use, production, trafficking, and consumption are also limited. This has been attributed to a lack of capacity for data collection and analysis [64]. In Egypt, there are reports that substance abuse rates are as high as 10%-20% the global average, with cannabis and tramadol being the most abused substances [65]. In Tunisia, an increase in the trafficking and consumption of psychoactive substances have been observed since the political uprising that occurred in 2011[66]. There have also been reports of increased availability of drugs of abuse, particularly to school students[66]. These increases have been confirmed by a few epidemiological studies[66-68]. Moslah et al[68] carried out a study to examine the pattern of substance use among 298 persons with a history of drug use between 2010 and 2015. The results showed that among these cohort of young adults, cannabis was the most widely consumed illicit drug, followed by benzodiazepines, buprenorphine, cocaine, and ecstasy [68]. Reports from the Mediterranean School Survey Project on Alcohol and Other Drugs (II) carried out in Tunisia in 2017 revealed that tobacco, alcohol, and cannabis were the substances most frequently used by secondary school students^[67], whereas psychotropic drugs such as ecstasy, cocaine, and buprenorphine were less frequently consumed. More importantly, it was observed that the frequency of use of these substances has increased significantly since the first survey published in 2014[66]. Chaouali et al[69] carried out an epidemiologic/toxicological investigation to evaluate patterns of drug abuse in 11170 suspected drug users. Urine samples collected between January 2016 and December 2018 were also analyzed. Results revealed a preponderance of males (97.4%) compared to females, with a median age of 29 ± 7.91 years. Also observed was that a large percentage of these drug users were single (91.3%). Examination of the urine samples revealed that about 48.4% tested positive for illicit drugs, with cannabis being the most widely consumed drug (95%), others were benzodiazepines, buprenorphine, cocaine, and opiates (0.13%). There was also a history of poly drug use[69]

In Southern Africa (although there are limited national data in most countries in the region), a rise in substance use has been reported [70]. In some of these countries including Zimbabwe, there is anecdotal evidence suggesting an increase in substance use among adolescents and young adults, with prevalence of substance use reportedly ranging from 6.1% to 13.8% [70]. Alcohol, cannabis, heroin, glue, and cough mixtures are among the most commonly consumed products in Zimbabwe. Cannabis, which is commonly known as mbanje, is grown locally (also smuggled into Zimbabwe from Malawi and Mozambique), and remains the most popular illicit drug among young Zimbabweans. Drugs are also trafficked through Zimbabwe to other countries in the region, including South Africa. In South Africa, an increase in substance use has been reported, which has been linked to the increased availability of illicit drugs including cannabis, cocaine, heroin, amphetamines, and ecstasy; either from diversion during trafficking or increased cultivation and local production [29,30,71,72]. Other factors that have contributed to the increase in substance use include an increase in migration and easing of border controls following the commencement of democracy, which have facilitated the development of youths'



movements that indirectly or directly promote substance use[30]. Results from surveys have revealed a gradual increase in cannabis consumption among adolescents and young adults in South Africa.

A 2002 school-based survey reported that 13% of the students (aged 19 years and below) had an history of cannabis use, although current use was 9%. About 12% had a current use of heroin, 11% used inhalants, and 6% consumed Mandrax^[73]. The results of another study (a 2005 National household survey) showed that the prevalence in the past 3 mo for cannabis among 15-19-years-old was 3% [74]. In another study examining the prevalence and patterns of use of illicit substances among persons presenting at drug treatment centers in South Africa, it was revealed that cannabis (16.9%), methamphetamine (12.8%), cocaine (9.6%), and prescription drugs (2.6%) were the substances commonly used among patients. Also, there was evidence of poly drug use, with cannabis and mandrax having a prevalence of 3.4%, whereas heroin and opiates had a prevalence of 9.2% [75].

Prior to 1994 and the first democratic elections, alcohol, cannabis, and methaqualone were the primary substances of misuse in South Africa. With South Africa's transition to democracy and subsequent reopening of borders, there has been an influx of and a growing burden of harm associated with illicit drug use. Alcohol, however, remains the most commonly misused substance, with 14% of the population having a lifetime diagnosis of alcohol abuse and/or dependence (Herman et al[76], 2009). Although the overall levels of alcohol consumption do not exceed those in the developed world, the pattern of consumption differs markedly, with hazardous and binge drinking being common.

New and emerging psychoactive substances in Africa

Use of novel psychoactive substances is an emerging trend in substance use that is fast becoming a public health challenge globally [77,78]. Novel or new psychoactive substances have been defined by the United Nations Office on Drugs and Crime^[79] as substances of abuse (existing either in its pure form or as a preparation) that are not controlled by either the 1961 or 1971 conventions on narcotic drugs and psychotropic substances, respectively, but pose significant threats to public health globally due to spikes in intoxications and fatalities associated with their use[79,80]. The term 'novel' or 'new' that is used in relation to these substances depicts their recent emergence in the global market. Substances that currently fall within the novel psychoactive substance category include (but are not limited to) synthetic cathinone and cannabinoids, synthetic opioids, image and performance-enhancing substances, tryptamine derivatives, piperazines, phencyclidine-like dissociatives, gamma amino-butyric acid (A)/beta receptor agonists, novel hallucinogens, benzodiazepines and psychotropic plants/herbs[77,81, 82].

In the last few years or more, there has been a growing demand and supply chain for these new psychoactive substances [77,78,81]. In the last 10-12 years, the number of novel psychoactive substances has increased considerably. In 2009, only about 166 of them had been detected; however, by 2019, the number had risen to about 950, with more than 70% of these substances available in Europe[80,83]. While in the developed economies, a lot is being done to ensure continued documentation of novel psychoactive substances as they emerge, it would seem that Africa is only beginning to awaken to the emerging public health threat that these substances pose to her teeming population of adolescents and young adults[82]. While the lifetime prevalence of novel psychoactive substance use in countries such as the United States have been examined [80], there is a paucity of data on the prevalence of novel psychoactive substance use in Africa. Although across the continent, there is increasing awareness of the dangers of novel psychoactive substance use.

In 2017, attention was called to an increase in the use of designer drugs in Nigeria. Some of these substances which have street names such as "black mamba", Colorado", "Lamba", "happy boy", and" Scooby snax" are believed to contain synthetic cannabinoids. Their use is associated with a rise in the incidence of hallucinations, convulsions, psychiatric disorders, kidney failure and fatalities[84]. News outlets, including the British Broadcasting Corporation News and Premium Times, also reported that the use of and addiction to non-conventional psychoactive substances such as tramadol and codeine cough syrups among Nigerian youths was reaching epidemic proportions[85,86]. In Nigeria, available novel psychoactive substances also include mixtures with street names such as "gutter water", a cocktail of cannabis, tramadol, codeine and ethanol), and "monkey tail", a cocktail of locally made gin and cannabis (seeds, leaves, stems, and roots). Some people have also been in a state of euphoria from drinking the mixture of specific carbonated drinks and menthol flavored candies[82]. The sniffing of dry human fecal matter, dry cassava leaves and seeds, Datura metal seeds, Moringa leaf, burnt tires, sewer gas, and nail polish have also been reported [82,87]. Different parts of some lizards, including the whitish part of their dung, are also smoked in a bid to achieve a "high" [82,87]. The inhalation of urine, sewage, petrol, and glue are also common practice among drug users in Africa. It is believed that the hallucinogens present in hydrocarbons from petrol, and gases produced from fermentation of sewage have the ability to cause a "euphoric high" similar to (but longer lasting) when compared to that derived from the ingestion of cocaine[88].

In southern Africa, particularly South Africa, there have been reports of the use of "Nyaope" also known as "Whoonga", which is a cocktail of low-grade heroin (black tar heroin), marijuana, antiretroviral drugs (Efavirenz), and other undisclosed substances [88]. In East African countries such as Uganda and Kenya, the habit of using novel psychoactive substances such as the sniffing of aviation gas/jet fuel, toluene and glue is reaching epidemic proportions among persons aged between 16-25



years[89-91]. Other substances that are abused in this region include "kuber" and "shisha", also known as hookahs ,which are variants of smokeless tobacco. There have been reports that compared to cigarettes, the smoking of the shisha or hookah pipe exposes the user to higher volumes of smoke containing high levels of benzene, tar, and other carcinogens and increased risk of lung cancer[92-95].

In Northern Africa, the smoking of hashish and the chewing of Kath has become very rampant. Although there is little data from the region regarding the prevalence and patterns of use of novel psychoactive substances; reports from studies carried out in Egypt have suggested that the estimated prevalence of novel psychoactive substance use among adolescents in the country are largely underestimated [96]. However, tabloid reports have called attention to an increasing demand and use of novel psychoactive substances including "voodoo" and "strox" among adolescents and young adult in Egypt [97,98]. "Voodoo" is gaining popularity rapidly, and it is usually packaged and sold as an herbal incense. "Voodoo" is a heterogeneous mixture of several psychoactive substances, including synthetic cannabinoids, tramadol, amphetamine, methadone, benzodiazepines, penitrem A (a neurotoxin) and morphine derivatives. The concentrations of the chemical constituents and adulterants of voodoo also vary substantially among the different clandestine laboratories that produce it[96,99]. Another novel psychoactive substance that is gaining popularity in Egypt is "Strox" [100]. "Strox" or "Egyptian Spice" has been reported to account for approximately 4.3% of the over 10400 patients requiring medical support for drug-related complications[101]. Also, addiction to "strox" was responsible for 22% of calls to the addiction center hotline[101]. "Strox" is a potent synthetic narcotic that is mixed with tobacco and smoked; it is compounded in clandestine laboratories by adding veterinary grade chemicals to aromatic herbs such as marjoram. There have also been reports of the addition of pesticides to increase the potency, although this increases the toxicity [97].

The search for novel psychoactive substances is fueled by the need to create drugs that are able to evade the chemical processes used for detection and the legal processes that criminalizes the use and possession of conventional drugs of abuse. Also, the need for compounds that deliver fast and sustained psychoactive effects when compared to the conventional drugs also drive the search for novel psychoactive compounds[102]. However, the variability of the chemical constituents and/or adulterants of the different compounds present a conundrum for the health professional who has to decipher and manage the divergent symptoms and signs that complicate the use of these substances. Hence, there is an increasing need for continuous surveillance so that new or emerging psychoactive substances can be discovered before they wreak havoc on our communities.

African plants and herbs with psychostimulant potential: Are they being abused?

Several plants and parts of plants have been shown to have central nervous system effects[103-109]. Also, current literature reveals that novel psychoactive substances can be derived from either synthetic compounds or from bioactive principles of natural compounds. These bioactive principles which are mainly alkaloids are present in a wide variety of plants including Ayahuasca, Catha edulis and nicotiana tabacum; and have been reported to possess hallucinogenic and/or stimulant effects[110]. Plants with psychoactive properties are found all over the globe and have been used for centuries by humans, for religious, therapeutic and recreational purpose[111,112]. Studies have shown that the bioactive principles of these plants enable the profound alteration of the human perception allowing for divination, ancestral contact, and spiritual enlightenment[111-113].

Africa has a high floral diversity and a rich tradition of indigenous medicinal plant and herb use [111]. Africa is also rich in flora of medicinal plants that possess central nervous system effects[107]. Although there is a paucity of ethnobotanical surveys on African plants with psychoactive effects, evidence from African traditional healers and diviners who use plants such as the 'Ubulawu', a preparation containing Sileneundulata and Synaptolepis are pointers that there are plants indigenous to Africa that contain compound which have mood altering effects[113]. A few plant species that are indigenous to Africa, such as the Cola species (Cola nitida, and Cola acuminata), Catha edulis (Kath), Datura species (Datura stramonium), Pausinystalia yohimbe (Burantashi Pausinytalia yohimbe) and Tabernanthe iboga have reported psychoactive properties[111,114], and have been used recreationally (Table 2) for centuries in the countries in which they are cultivated. However, in recent times, the use of and dependence on some of these plants by adolescents and young adults (either alone or combined with established illicit drugs) is reaching epidemic proportions. In this section, we reviewed the abuse potential of some psychoactive plants that are indigenous to the African continent.

Kath (Catha edulis Forsk) is a flowering plant native to countries in East Africa and the Horn of Africa. Fresh young leaves and twigs from Kath are chewed daily by large populations of people for its psycho-stimulatory properties. The chewing of Kath dates back centuries, being a practice that is rooted in tradition, social custom and the culture of the indigenous populations[115]. It has been reported that more than 20 million people worldwide chew Khat[116,117]. Although traditionally a custom associated with older middle Eastern and Eastern African men, Khat's use is now expanding to include women and younger persons. In the Eastern region of Ethiopia, approximately 30% of adolescent girls and 70% of adolescent boys chew Khat. The active principle contained in Khat is cathinone (an alkaloid), which is a stimulant that causes excitement, appetite loss and euphoria[117]. In countries such as Somalia, Ethiopia, Djibouti, and Kenya, the dependence on Kath is warranting its consideration as a substance of abuse. In Somalia a law prohibiting the use, cultivation, importation and trade of Kath was enacted and



Table 2 African plants and herbs with psychostimulant potential

Region	Herbal preparation	Plant	Bioactive compound	Central nervous system activity	Toxicity	Ref.
East Africa	Khat chewing, drink made from dried leaves or smoking dried leaves	Catha Edulis	Phenylalkylamines and the cathedulins (Cathinone)	Improves performance, stay alert and to increase work capacity, excitement, appetite loss and euphoria	Memory impairment, sleeping disorders, liver toxicity, cardiovascular disease, psychosis and poor academic performance	[115, 117, 124- 126]
West Africa	Different parts of the plant are smoked or used to make concoctions	Datura specie including stramonium and Datura metal	Atropine, scopolamine, and hyoscamine	Anticholinergic and hallucinogenic activity	Hyperthermia, tachycardia, delirium, pronounced amnesia, severe mydriasis, bizarre behaviors and painful photophobia	[135, 136, 141, 145- 148]
West Africa	Root bark concoctions	Tabernanthe iboga	Ibogaine	Stimulatory, hallucinogenic, and sedative effects	Development of ataxia, tremor, cardiac toxicity, and death	[149- 151]
South Africa	Ubulawu drink	Silene undulata and Synaptolepis	Triterpenoid saponins	Mood altering effects including stimulating vivid or lucid dreams	Confusion	[113]
South Africa	Chewed, smoked, snorted or swallowed	Sceletium tortuosum	Mesembrenone, mesembrenol, mesembrine and tortuosamine	Increased libido, decreased stress, euphoria and appetite suppression	Anxiety, headache, hypertension, irritability, insomnia and nausea	[154, 155]

enforced by comprehensive national program [118]. At about the same period (approximately two decades ago), Kath was also considered by the World Health Organization and classified as a drug of abuse, although its abuse potential was not thought to constitute a serious problem compared with that of alcohol or tobacco[119]. Across a region extending across Africa and the Middle East, predominantly among the Ethiopians, Somalians and Yemenis, approximately 5 to 20 million people use Kath[116,120], with the consumers engaged in the practice for the best part of a day resulting in a loss of manpower and national income [121,122]. In 2005, a survey by the World Health Organization revealed a prevalence of 20% Kath abuse in Kenya, exceeding the prevalence observed in most of the other countries in the region [123]. However, more recent studies are demonstrating that in spite of attempts by the respective countries to criminalize the use of Kath, Kath chewing is fast becoming a common practice among young adults in countries like Kenya, Ethiopia, Somalia, Djibouti[124-126]. In Ethiopia, a study carried out among academic staff of a university revealed that the lifetime prevalence of Kath chewing was 41% [126], while another study carried out among college students reported a prevalence of 42% [124]. In Kenya, a recent household survey revealed that the prevalence of current Kath chewing in the region was 36.8% [125], which would suggest a significant rise from the 20% prevalence reported by the World Health Organization[123]. While Kath chewing was not previously known outside the regions within which it was cultivated, the effects of migration and trade have propelled it to a widely used psychostimulant globally[115,127]. Kath chewing has been associated with adverse health effects that is creating public health challenges in countries across Asia, Europe, Australia, and the United States of America[128-130]. When chewed concurrently with tobacco, there have been reports of cardiovascular stress response[129]. It has also been associated with the alteration of physical, mental, social and cognitive aspects of human functioning[131]. Chewing Kath chronically has been reported to cause memory impairment, sleeping disorders, liver toxicity, cardiovascular disease, psychosis and poor academic performance[126,132]. While attempts are being made by countries to criminalize the importation and trade of Kath, smugglers continue to find new avenues and trade routes. For example, in 2016, the National Drug Law Enforcement Agency of Nigeria reported seizures of Kath load, which was possibly destined for the Nigerian market or enroute countries[133]. However, in 2020, the United States customs reported seizure of Kath load from Nigeria destined for the United States suggesting that Nigeria is fast becoming a Kath transit hub[134].

In Nigeria, complicating the substance abuse epidemic is the emerging trend of experimenting with plant extracts or brews from a group of flowering plants belonging to the Datura specie, of the nightshade family Solanaceae [135,136]. Although members of the Datura specie which are broadly known as thorn apple, devil's apple, devil's trumpet or angel's trumpet have their origin in central Americas and in the south-west region of the United States if America[137,138], they have become naturalized all over the world, being widespread in Asia, Europe, and Africa[139]. The datura specie is made up of herbs and shrubs with erect or branched stems, with alternate simple basal leaves and opposite leaves on terminal branches[139]. The fruit has a spiny capsule and reniform seeds[139,140]. All parts of the Datura stramonium and Datura metal plant has been shown to contain tropane alkaloids such as atropine, scopolamine, and hyoscyamine which have significant anticholinergic and hallucinogenic activity[141]. The high tropane alkaloid content of these plants increases their medicinal value and also opens them up to potential abuse. In Nigeria, Datura stramonium (thorn apple, devil's

snare, devil's trumpet or jimsonweed) and Datura metal (Indian thorn apple) are naturalized. Similar to a number of countries across the world (United States of America and Canada) where there has been reports of datura-induced poisoning among adolescents who abuse the plant for its hallucinogenic effects[142-144], adolescents and young adults in Nigeria are also experimenting with the plant and getting poisoned[136]. Datura metal grows wildly (although at times it is cultivated) across the different regions of Nigeria where it is called 'Myaramuo' by the Igbos of south eastern Nigeria, 'Zakami' by Hausas of northern Nigeria and 'Apikan' by the Yorubas of southwestern Nigeria[145,146]. Datura stramonium also grows as a weed and is also cultivated across the different states of Nigeria. It is known as 'Gegemu' by the Yorubas and 'Zakami' by the Hausas[136]. Both plants have been reported to have hallucinogenic and euphoric effects when the different parts of the plant are either smoked or used to make concoctions[147]. Datura stramonium poisoning is associated with hyperthermia, tachycardia, delirium, pronounced amnesia, severe mydriasis, bizarre behaviors and painful photophobia[136,148]. These features can appear as early as 30 min to 1 h following consumption of the extract or smoking of the weed and have been reported to last several hours to days or at times even as long as 2 wk[148].

Another plant with psychoactive properties is the Western African shrub Tabernanthe iboga from whose root bark ibogaine, a hallucinogenic alkaloid is extracted. Traditionally, the concoctions from the roots have been used for their stimulatory, hallucinogenic, and sedative effects[149]. Ibogaine has been reported to exhibit stimulant effects at low doses and result in the development of hallucinations at high doses. Its use has also been associated with the development of ataxia, tremor, cardiac toxicity, and death[149-151]. There have also been reports that ibogaine has anti-addictive properties, although its use is limited by its deleterious effects[152,153].

In southern Africa, the use of extracts, dried-powdered herb, tincture, tea bags and seeds of the plant sceletium tortuosum also known as Kanna is also gaining traction. These different compositions of the plant can be chewed, smoked, snorted or swallowed resulting in increased libido, decreased stress, euphoria and appetite suppression. There are reports attributing the antidepressant and mood-elevating effects of the plant to the serotoninergic activity of its alkaloids including mesembrenoe, mesembrenol, mesembrine, and tortuosamine. Indiscriminate use has been associated with the development of anxiety, headache, hypertension, irritability, insomnia and nausea[154,155]. A serotonin syndrome has also been observed especially when consumed alongside selective serotonin reuptake inhibitors or monoamine oxidase inhibitors. Although the use of a number of these plants and herbs may not be illegal in the countries in which they are consumed, increasing reports of poisoning arising from the use or these psychoactive plants solely or in combination with other compounds is drawing attention to the need to enact public health laws that can criminalize their use.

Synthetic cannabinoid in herbal products

Synthetic cannabinoids are compounds which are structurally similar to natural cannabinoids [tetrahydrocannabinol and cannabidiol (CBD)] allowing them to exert their effect through binding to cannabinoid receptors (CBD1 and CBD2)[156,157]. Synthetic cannabinoids can be agonists at the CB₁ receptors or antagonists at other cannabinoid receptors. Although many of the synthetic cannabinoids are used in pharmacology in structure - activity relationships and receptor binding studies, others have medicinal uses including in the treatment of anorexia, as antiemetics in cancer chemotherapy and in pain management. In the last two decades, commercial preparations containing synthetic cannabinoids have become popular for their use as designer drugs marketed as herbal incense or herbal blends under the names 'Spice', 'synthetic marijuana', and 'K2'[158-160]. The cannabinoid compound is sprayed onto inert plant material and smoked or ingested in liquid form[160,161]. Although often considered legal and safe alternatives to cannabis, there is evidence indicating that synthetic cannabinoids use is associated with significant health risks when compared to marijuana; there have also been reports that their distinct pharmacological effects and metabolic activity could also be a contributing factor to the increased toxicity observed following their use[157,162,163].

To date, the abuse of herbal preparations that have been spiked with synthetic cannabinoids continues to increase. This is evidenced by an increasing list of commercial preparations marketed in the United States and Europe under the names fairly legal, Pandora's box, Angry birds, exodus, bonzai, annihilation, weekend blend, fire, strong spice, green Buddha, smoke, and Scooby snacks[102,164]. In the last few years, Africa is also beginning to experience a surge in demand for and use of synthetic cannabinoids. In Mauritius, since the year 2015, there has been a reported increase the number of arrests involving synthetic cannabinoids[165]. In different countries in the continent they are marketed under various street names including, Wiz in South Africa[166]. In Nigeria it is marketed as Black Mamba, Colorado, Lamba, Happy Boy or Scooby Snax[167].

Across Africa, available evidence points to a growing use of novel psychoactive compounds which mainly contain synthetic cannabinoids. Synthetic cannabinoids have effects that are similar to that experienced with natural cannabis, although they are more potent and have been associated with more severe physical and psychological adverse effects necessitating hospitalizations[168,169].

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PREVALENCE OF SUBSTANCE USE DISORDERS AND AVAILABILITY OF EVIDENCE-**BASED TREATMENT CENTRES IN AFRICA**

Substance use disorders are defined as the persistent use of alcohol or other psychoactive substances despite significant harm and untoward health consequences [170]. They are characterized by an array of social, emotional and behavioral problems. Across Africa, there is also a dearth of scientific data on the prevalence of substance use disorders or drug dependence [48,55,56,171]. In Nigeria, reports from the 2018 National Drug Survey revealed that one in every five persons who used drug in the past year also had a drug-related disorder[48,55,56]. In South Africa, results obtained from a nationally representative sample of 4351 persons aged 18 years and above revealed a lifetime prevalence of substance use disorders of 13%, with alcohol use disorder being the most prevalent type of substance use disorders[72, 171-173]. In Egypt, reports obtained from the National Addiction Research Study revealed the prevalence of drug dependence in the different regions ranged from 3.2%-9.3% [174].

Left untreated, substance use disorders contribute significantly to the global burden of disease, including increasing morbidity and mortality and societal cost implications such as increased healthcare costs, lost productivity and costs related to social welfare and criminal justice[4,5]. Access to evidencebased treatment has been linked with a reduction in the risk for ill health[171]. Accordingly, towards reducing the burden of substance use globally, availability and access to evidence-based treatment facility were included in the United Nations' Sustainable Development Goals for 2030[175]. However, despite reports of increasing prevalence of drug dependence and substance use disorders, reports from surveys carried out in a number of countries in Africa suggest that the availability of treatment centers are limited [171,176]. Factors contributing to this treatment gap include treatment infrastructure constraints, poor funding, and the high cost of private-for-profit treatment centers[171,176].

How can Africa's burgeoning substance use and substance use disorder problem be addressed

It has become evident that there is a burgeoning illicit drug use problem across the African continent[47-49,177]. There had been predictions that by the year 2050, increased life expectancy and a rapidly growing population would result in approximately 130% increase in the burden of mental and substance use disorders to about 45 million years lived with disability in Africa[178]. While different factors, including increased access to illicit drugs and high level of youth unemployment have been adduced to explain the emerging drug use pandemic; its significant contribution to economic instability, crime, criminality and insecurity across Africa and worldwide means that governments and policy makers need to prioritize the need to develop ways to mitigate these problems. The dearth of comprehensive data and the uniqueness of the manifestation of illicit drug use to individual countries within the African region are factors that impede progress towards addressing this looming pandemic.

Understanding the different determinants of drug use within the different populations of Africans and how these impact the prevention and treatment of substance abuse disorders in the individual countries would be an important step towards addressing this emerging pandemic. Currently available data suggests that influencers of drug use (particularly in adolescents) which include family, social networks and peer pressure are common to most of the countries [179-184]. Other determinants of drug use also include childhood trauma and adverse life experiences such as sexual, emotional or physical abuse. Across age groups, demographic factors such as being male, lower level of education and attendance of private schools have also been reported by researchers from the different regions of Africa [185-187].

In addition to understanding the influencers and determinants of drug use, there is also a need for up-to-date national and regional data that can adequately determine the prevalence and incidence of drug use across all demographics. The availability of a detailed and comprehensive national data would provide a background against which policy successes or failures can be measured, it would also alert governments and international partners on the need for increased funding or more treatment facilities.

The deleterious health effects of drug use disorders means that the provision of effective prevention, treatment and care facilities for substance use disorders is a necessary investment in the health of the society as a whole. Research has shown that the availability of evidence based prevention programs and policies have the ability to significantly reduce substance use and related harmful effects[188]. Behavioral and medication-assisted treatment using a chronic-illness-management approach has also been shown to aid recovery and prevent relapse. There have been suggestions that easy access to support services assist previous substance users to achieve and maintain wellness long-term[188].

Addressing the ease of access to drugs and other illicit substances within communities and regions need to be taken more seriously. There is a need to gather information on the different types of novel psychoactive substances that are available within communities and also create awareness as to the adverse health effects associated with consuming these compounds.

Limitations and recommendations

One of the major limitations encountered in this review was the dearth of recent, community based and age specific scientific data on the prevalence and extent of the substance abuse problems in most of the countries in Africa. There was also a deficit of data on the details and impact of the country-specific



intervention protocols. This led to reliance mainly on third party data from international; partners and a few independent researchers. The battle to win this emerging substance-use pandemic in Africa can only be successful if there is increased emphasis in documenting the extent of the problem and country specific interventions; particularly at the community levels with emphasis on how different age groups are impacted by substance abuse.

CONCLUSION

In Africa, substance use and substance use disorders drain struggling economies and health care systems. While the continent might have some general idea of what it is up against, understanding the details of the problem and availability of the willpower/wherewithal to subdue it remains a challenge. It is becoming obvious that there is no substitute for well-designed, accurate and comprehensive population-focused efforts at obtaining data that relates to substance use and substance use disorders, since such data will form the foundations for designing effective intervention strategies. Also, in Africa, interventional strategies should place emphasis on prevention, through identification of and mitigation of risk factors, as this approach is likely to consume less resources in the long run.

FOOTNOTES

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MINIREVIEWS

Artificial intelligence-assisted psychosis risk screening in adolescents: Practices and challenges

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Abstract

Artificial intelligence-based technologies are gradually being applied to psychiatric research and practice. This paper reviews the primary literature concerning artificial intelligence-assisted psychosis risk screening in adolescents. In terms of the practice of psychosis risk screening, the application of two artificial intelligence-assisted screening methods, chatbot and large-scale social media data analysis, is summarized in detail. Regarding the challenges of psychiatric risk screening, ethical issues constitute the first challenge of psychiatric risk screening through artificial intelligence, which must comply with the four biomedical ethical principles of respect for autonomy, nonmaleficence, beneficence and impartiality such that the development of artificial intelligence can meet the moral and ethical requirements of human beings. By reviewing the pertinent literature concerning current artificial intelligence-assisted adolescent psychosis risk screens, we propose that assuming they meet ethical requirements, there are three directions worth considering in the future development of artificial intelligenceassisted psychosis risk screening in adolescents as follows: nonperceptual realtime artificial intelligence-assisted screening, further reducing the cost of artificial intelligence-assisted screening, and improving the ease of use of artificial intelligence-assisted screening techniques and tools.

Key Words: Psychosis risk; Adolescents; Artificial intelligence; Big data; Social media; Medical ethics; Chatbot; Machine learning

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Core Tip: Artificial intelligence-assisted psychosis risk screening must be emphasized and applied in adolescents. This review summarizes the application of two artificial intelligence-assisted screening methods (chatbot and large-scale social media data analysis), and proposes that the first challenge in applying artificial intelligence to psychosis risk screening concerns ethical issues. The methods must follow four biomedical ethics principles, *i.e.*, respect for autonomy, nonmaleficence, beneficence, and justice. Three directions should be considered in the future: nonperceptual real-time artificial intelligenceassisted screening, further reducing the cost of artificial intelligence-assisted screening, and improving the ease of use of artificial intelligence-assisted screening techniques and tools.

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INTRODUCTION

In recent years, the prevalence of psychosis among adolescents has been increasing. According to the data released by the World Health Organization, approximately one in five children and adolescents worldwide suffers from a mental disorder, and half of these individuals show symptoms before the age of 14[1]. The risk of psychosis shows not only a trend of a younger age but also many disease categories and high heterogeneity[2-5], and its potentially high prevalence warrants attention. Two studies in Europe and North America revealed that adolescents with prodromal symptoms of psychosis who actively sought help still experienced a risk of eventual psychosis at rates of 19% and 35%, respectively [6,7]. In fact, due to the lack of independence, fear of being discriminated against by people around them, and a dearth of adequate attention from parents and schools, some adolescents even do not actively seek medical treatment and thus often miss their best treatment window. Since 2020, the rapid spread and persistence of the coronavirus disease 2019 pandemic worldwide has caused further major and hidden risks to population health, including psychosis. A study revealed that the deadly hazard of coronavirus disease 2019 and the resulting national lockdown policies in Italy caused intense psychosocial stress in individuals, which can be a trigger for first-episode psychosis[8]. In particular, children and adolescents are the most vulnerable groups[9]. Compared with other groups, children and adolescents are more vulnerable to the negative effects of the pandemic[10,11], e.g., anxiety, depression, and posttraumatic symptoms, which increase the psychosis risk and may cause long-term negative consequences [12,13]. Additionally, adolescents are at a greater risk of first-episode psychosis than adultonset psychosis, which is often associated with more severe symptoms and worse outcomes [14,15]. People who suffer from severe mental disorders die 10 years to 20 years earlier than the average person [1] and are more vulnerable to long-term disadvantages in terms of career advancement in the labor market, social status, mental health, and life beliefs [2,16]. Undoubtedly, this reality is not conducive to the stable economic and social development of any country worldwide.

Influencing factors

The prior literature has concluded that psychosis risk is the result of an interaction between various internal and external factors[12,17], which can be divided into three main categories. First, national and social factors, such as those associated with schizophrenia, are unequally distributed across cultures and countries^[18], and a cultural atmosphere that stigmatizes psychosis can create barriers to the timely detection of the illness^[19]. The second category comprises family factors, including the two aspects of congenital inheritance and acquired growth environment. Adolescents who grow up in families with psychotic parents, domestic violence, or abuse are at a greater risk of psychosis[20]; adolescents suffering from psychosis have higher rates of broken homes, substance abuse, chronic disease[21], and lack of social capital^[22]. Third, individual factors include demographic characteristics and addictive behaviors; for example, males account for a larger proportion of children and adolescents with firstepisode psychosis[21], while marijuana use is also a risk factor for psychosis[23,24]. Indeed, a combination of risk factors, including genetics, birth season, birth complications, infection and immune system factors, autoimmune diseases, ethnicity, marijuana use, and urban residence, increase an individual's risk of developing schizophrenia[18].

Main benefits of artificial intelligence-assisted adolescent psychosis risk screening

Despite this severe reality, shortages of medical resources and professional psychiatrists and uneven medical care are still common. Many nonpsychiatric specialists in hospitals and psychological service personnel on campus or in the community are unable to accurately and efficiently identify psychotic patients, and even if they diagnose the condition, they still cannot perform effective follow-up and



treatment[25-28]. A consensus holds that one of the best strategies to promote early intervention against psychosis is to improve the early identification of individuals at risk for psychosis through screening[29-31]. However, screening for psychosis mainly relies on scales, complicating the accurate identification of adolescents with a psychosis risk in a timely manner.

Over the past few years, artificial intelligence has shown explosive development inseparable from the emergence of new algorithms and the speed of high-performance parallel computing, coupled with the development of large-capacity storage space and video, text, sound, and other technologies to promote its rapid growth. Recent advancements in artificial intelligence have promoted improvement in the methods and technological innovations used in the treatment of human mental diseases, and artificial intelligence-based technologies are gradually being applied to psychiatric research and practice [19,32-35]. The main benefits of artificial intelligence-assisted adolescent psychosis risk screening are as follows: (1) Compared with traditional screening, the introduction of artificial intelligence can improve the speed and timeliness of identifying those who are already sick or who have a potential risk of a disease[36], which helps with early intervention and treatment[37,38] and timely correction of patients' risky behaviors, all of which can prevent the occurrence and further aggravation of symptoms; (2) Using advanced technology and objective data, artificial intelligence further enhances the accuracy and objectivity of screening methods. Appropriate screening tools that have been developed for all conditions in adolescent psychosis risk are relatively inadequate [39,40], while the clinical significance of adolescent self-assessment results is limited[41]; and (3) Artificial intelligence mitigates the scarcity of medical resources[42] and increases the coverage of screenings. Additionally, artificial intelligence can process massive amounts of data and use these data to improve generalization[43,44] while playing a pivotal role in identifying and detecting heterogeneity in schizophrenia and other mental illnesses[5] and can help doctors make the right decisions for subsequent diagnostic treatment[45].

Existing challenges

Ethical issues constitute one of the greatest challenges encountered in the application of AI to psychosis risk screening in adolescents in terms of both technical development and concrete practice [46,37] in the following four aspects: (1) Whether the autonomy of adolescents to participate in screening is duly respected and protected [47]; if the screening is conducted without their full approval, they should be responsible for the possible negative consequences; (2) The personal information and privacy of adolescents are leaked and exposed to unauthorized surveillance and security risks; the use and management of data collected based on artificial intelligence technology deserves attention [48,49]; (3) There is no unified understanding of the ethical assessment and acceptance of technology among different stakeholders[37]; and (4) The benefits of AI technology development do not reach all adolescents fairly and equitably.

Contribution

Clearly, there is a strong necessity and feasibility to focus on and apply artificial intelligence-assisted psychiatric risk screening in adolescents. However, there is a paucity of research concerning artificial intelligence-assisted psychiatric screening and a dearth of narrative literature reviews focusing on this important population characteristic of adolescents. Therefore, this paper reviews the main literature concerning artificial intelligence-assisted adolescent psychiatric risk screening to clarify the current state of development and recent explorations of this important topic in terms of practice and challenges with the aim to contribute to a more effective use of artificial intelligence methods for adolescent psychiatric risk screening in the future on a global scale.

PRACTICES

Traditional psychosis risk screening methods are mostly based on various self-assessment questionnaires with obvious limitations as follows: (1) Performance is not comparable among different screening tools; (2) The measurement criteria (such as content, number of items, and thresholds) widely vary; and (3) Dynamic and longitudinal tracking data are lacking[40]. In addition, scale-based self-assessment relies on individual self-perceptions, recollections, and subjective evaluations, and in some situations, individuals may exaggerate or mask some of their symptoms, weakening the accuracy of the results. For example, a general recall bias is evident among patients with depression, and symptoms can fluctuate over time or even throughout the day, which complicates capturing dynamic changes in symptoms with high accuracy [50].

The emergence of artificial intelligence can address and largely overcome the above limitations. The main machine learning algorithms currently used for psychosis screening are traditional ones, e.g., decision tree, naive Bayes, random forest, support vector machine, K-nearest neighbor, and shallow neural networks. Of these, relevant studies have shown that the support vector machine method is the most commonly used[51,52]. With the advancement of deep learning algorithms, algorithms, such as convolutional neural networks, autoencoders, and deep belief networks, have begun to be used in psychosis risk screening research and are viewed as an important development trend of the future [3,53,



54]. By summarizing the pertinent literature, the artificial intelligence tools most often applied to psychosis risk screening are chatbot and large-scale social media data analysis.

Chatbot

Chatbot is a computer program that allows human-computer interactions in the form of textual dialog based on the technology of natural language processing[55]. The world's first chatbot, ELIZA, was developed in the 1960s[56] and responds according to special rules by recognizing keywords in userentered texts[57]. Due to substantial advancements in artificial intelligence, chatbots have developed from being driven by static databases and learning new responses and contexts based on real-time interactions with humans to the fusion of real-time learning and evolutionary algorithms. Currently, chatbots have powerful capabilities of simulating the structures of natural language communication and creating a realistic environment in which users can achieve human-computer interaction. Chatbots in the healthcare field include Tess, HealthBuddy, Florence, Buoy Health, and Your.Md. In addition to natural language processing, the machine learning methods adopted by chatbots also include natural language understanding, artificial neural networks, and recurrent neural networks[58].

The prior literature has shown that psychosis is usually strongly correlated with human manifestations, such as facial expressions, voice, textual tone, and gestures. According to these human manifestations, chatbots with cognitive ability can ascertain the needs of users in real time to provide emotional responses and predictions and assessments of their mental health conditions [46,59]. Based on existing experience, one study improved upon the feature extraction of previous studies by using deep learning and fusion regression methods to construct an artificial intelligence system that automatically predicted depression levels based on vocal and visual expressions, which showed better predictive performance than other existing methods using the same dataset. Artificial intelligence is currently used in some chatbots. This study used deep learning methods to extract key visual features from facial expression frames, spectral low-level descriptors and mel-frequency cepstral coefficient features from short audio segments, and time movements in feature space through feature dynamic history histograms (FDHHs). Finally, regression techniques were used to fuse these FDHHs and audio features to predict the Baker Depression Scale II scores. The artificial intelligence developed in that study was a general framework that can be used to automatically predict depression scale scores from facial and vocal representations. It has FDHH dynamic functionality, leveraging the ideas of motion history histograms on deep learning images and handcrafted feature spaces, and enables feature fusion of different descriptors of face images[34].

The chatbot Woebot is used as an example. Woebot can be used on mobile communication devices in the form of short daily conversations and mood tracking to help users acquire anxiety reduction skills by identifying cognitive distortions to monitor anxiety and depression episodes while using fully automated conversational agents to address poor adherence to some extent. In a previous randomized controlled trial using Woebot, 70 college students who reported symptoms of depression and anxiety were randomly assigned to an intervention group that chatted with Woebot in an instant messaging application and a control group that received the National Institute of Mental Health e-book on depression in college students. The results revealed that the anxiety levels decreased in both groups, and the students who interacted with Woebot had significantly lower levels of depression compared to those reading the e-book. Future validation of the findings is needed with more participants, longer doses, and longer-term follow-up data[60].

In summary, the advantage of chatbots is that they can bring hope to psychosis risk screening for those who were previously inaccessible to screenings or who are economically constrained^[61], build trusting relationships with potential patients, increase self-disclosure, and reduce the shame that patients or their families often feel when talking to doctors about mental illness. Nevertheless, these chatbots still have some shortcomings as follows: (1) They can be promoted by financial sponsors, causing conflicts of commercial interests; (2) In contrast to humans, they do not truly have subtle emotional awareness or empathic responses; and (3) They have issues with privacy, ethical risks, and other negative problems.

Large-scale social media data analysis

Currently, large numbers of users express their emotions and communicate daily through social media, such as Facebook and Twitter. Based on informative data such as textual information, emojis, user log information, and pictures, psychosis can be identified and predicted by combining natural language processing, sentiment analysis, and machine learning[49,62,63].

As the use of social media platforms becomes increasingly common in people's lives, screening for psychosis risk based on collected social media data will become easier. For instance, one study systematically analyzed artificial intelligence depression detectors and concluded that artificial intelligence systems that identify users at a high risk for depression from their social media data have made remarkable progress[37]. Given that depression is common in the adolescent population[26,64] and is underdiagnosed and undertreated, which underscores the need to expand the current screening methods, some investigators used the text of posts of consenting individuals on Facebook to predict depression as documented in electronic medical records and demonstrated correlative accuracy in identifying people with depression[65,66]. Therefore, the use of machine learning technology to screen



depression patients by acquiring the social media data for consenting individuals may become an effective and scalable supplement to existing screening methods [67]. Based on the language behaviors of Facebook user posts, Islam et al[68] achieved a classification accuracy of 99.0% with a depression prediction model using the decision tree method. Another study applied a logistic regression and highly randomized trees as modeling algorithms to approximately 20 million words of social media posts by 999 consenting volunteers and found that applying the method to Facebook posts significantly improved the predictive accuracy of demographic variables (age, sex, and ethnicity) in 18 of 21 disease categories, and it was particularly effective at predicting mental health conditions (anxiety, psychosis, and depression)[69]. In one study, big data were collected from China's Sina Weibo to understand differences in language style, emoji use, and the number of followers between depressive patients and nondepressed patients by using a deep neural network for feature extraction and dimensionality reduction. By constructing input data suitable for the classifier and applying the deep integrated support vector machine algorithm to classify the input data, the study achieved a more stable and accurate identification of depression in college students^[70]. The development of Internet-of-Things technology has realized the exchange of information between hardware such that various wearable devices can carry a large amount of health information. Applying the Internet of Things to the field of psychosis through machine learning, the objective behavioral characteristics collected through mobile phones and wearable devices can effectively predict depressive symptoms [71,72]. Data related to daily activity, sleep, social communication, etc. have been collected through smartphone sensors to predict individuals' depression situations [73,74]. Advanced artificial intelligence methods, including natural language analysis and chatbots, were used by the Horyzons website to analyze the sentiment and language of newsfeed posts and other relevant factors (e.g., user preferences and history), which enabled personalized treatment recommendations to be made for adolescents with early symptoms of psychosis [75]. Orabi et al[76] extracted unstructured text data posted on Twitter by 327 depression patients, 246 posttraumatic stress disorder patients, and 572 healthy individuals, and based on these data, users with depression tendencies were detected using the convolutional neural network method. Convolutional neural networks represent the most popular deep learning method in the field of natural language processing, boasting an accuracy as high as 87.9%, and have achieved remarkable progress in the field of image recognition[76]. In the future, more technologies, such as multimodal perception, understanding, and natural dialog and interaction (a multimodal auxiliary screening mechanism established through artificial intelligence perception technology), are needed to achieve more comprehensive and accurate screening of psychosis risk among adolescents.

CHALLENGES

The technology of artificial intelligence-assisted psychosis risk screening in adolescents will become more mature and a major development trend in the future. However, it can only do so by overcoming the existing challenges in the application of artificial intelligence-assisted psychosis screening in the adolescent population, which have rarely been addressed to date. Especially when applying artificial intelligence to psychosis risk screening, the primary challenge is ethical issues. The four principles of biomedical ethics, *i.e.*, respect for autonomy, nonmaleficence, beneficence, and justice[77], must always be firmly followed. On this basis, a new principle aiming to realize other principles through understandability and accountability^[78] such that the development of artificial intelligence can truly meet the moral and ethical requirements of mankind has been proposed. Table 1 presents the four widely accepted ethical principles, their connotations, and the corresponding issues that adolescents may face.

Respect for autonomy

Respect for autonomy requires respect for the patient's personal dignity and autonomy, such as ensuring informed consent and informed choice, ensuring that humans have complete and effective autonomy, and requiring that the operation of any artificial intelligence be supervised by humans. Adolescents are still minors, and this age group is in the typical age when psychosis develops. Discussions have addressed whether adolescents have autonomy and how they should be "empowered". For example, in the United Kingdom, adolescents under the age of 16 can be competent to give consent if they demonstrate sufficient maturity and intelligence (as judged through Gillick competence). For minors deemed incompetent (and adults who are incompetent due to mental illness), questions arise regarding whether guardian advocates or those with parental responsibility should be empowered to provide proxy consent for psychosis risk screening[80]. Therefore, assuming that a given artificial intelligence-assisted psychosis risk screening method is safe and trustworthy, improving the awareness and attitudes of teenagers and their parents toward the psychosis risk and the importance of early screening is vital as their willingness to use artificial intelligence for screening can be increased only with their full approval [79]. Moreover, how to improve adolescents' autonomous participation in psychosis risk screening by ensuring effective informed consent and meaningful disclosure of results still requires further discussion.

Table 1 Connotations of ethical principles and issues faced by adolescents				
Ethical principles	Connotations	Issues faced by adolescents		
Respect for autonomy	Ensuring informed consent and informed choice, ensuring that humans have complete and effective autonomy, and requiring that the operation of any artificial intelligence be supervised by humans	Safety and trustworthiness of screening methods; full approval from adolescents and parents; willingness to use artificial intelligence for screening[79]		
Nonmaleficence	Privacy, security and "capability warnings" [78]; artificial intelligence technology must be able to strongly resist malicious use, including avoiding harm to the natural environment and all living things	Privacy leakage and data abuse; difficulties in oversight and accountability; adverse effects and stigma with irreversible damage		
Beneficence	Must be beneficial for not only the patients but also the medical cause, medical sciences and even the well-being of the entire human race	Screening scales need to be refined; no consensus (such as ethical evaluation acceptance of the technology) among different stakeholders[37]		
Justice	Everyone in society has equal rights to reasonably enjoy health resources and participate in the distribution; prosperity is promoted; and unity is maintained	Development of artificial intelligence cannot benefit all groups of young people[12,14,21,22,80]; intergenerational transmission maintains inequality[46,78]		

Nonmaleficence

Nonmaleficence requires privacy, security, and "capability warning" [78]. To protect the integrity of the human body, mind, and dignity, artificial intelligence technology must be able to strongly resist malicious use, including avoiding harm to the natural environment and all living things. Smart data collection technologies are becoming increasingly powerful, posing a greater threat to user privacy and security. The protection of adolescents' personal information and privacy is very important, but some privacy leakage and data abuse problems remain, which have been extremely harmful. With the rapid development of artificial intelligence, the existing ethical and regulatory norms have fallen behind. Their failure to keep pace with the latest environmental and artificial intelligence technologies creates difficulties in oversight and accountability. Especially in the presence of potential commercial interests or vested interests, e.g., some social platforms may be abused by enterprises/people with criminal minds or illegal attempts, the usage of artificial intelligence in biomedical fields must be monitored and regulated from an ethical and moral standpoint. Furthermore, an artificial intelligence-assisted screening result may have adverse effects on some adolescents with a psychosis risk and introduce stigma when they are labeled with psychosis, which may, in turn, cause irreversible damage to their mental health, interpersonal relationships, and even long-term personal development. Thus, in addition to psychosis, many social factors related to its diagnosis cause extra damage to adolescents.

Beneficence

Beneficence requires that something be beneficial for not only the patients but also the medical cause, medical sciences and even the well-being of the entire human race. Although many studies have confirmed the positive role of artificial intelligence in psychosis risk screening, people still use scales for screening in practice. On the one hand, time is required to ensure that any technology is foolproof, and on the other hand, different stakeholders (adolescents and their parents, doctors, research and development personnel, *etc.*) have not yet reached a consensus, and the ethical evaluation and acceptance of the technology are still open questions[37], *e.g.*, whether medical professionals are willing to replace traditional screening with artificial intelligence-driven products and technologies. Therefore, whether this technology is truly beneficial for the health and well-being of society as a whole and humanity is worth discussing.

Justice

Justice requires everyone in society to have equal rights to reasonably enjoy health resources, resources to be fairly distributed, everyone to have the right to participate in the distribution and use of these resources and the benefits of artificial intelligence to be distributed fairly and equitably while avoiding any discrimination or stigma, promoting prosperity, and maintaining unity. The development of artificial intelligence cannot benefit all groups of young people. Previous studies revealed that adolescents born in the lower classes of society and those from disadvantaged families are more likely to suffer from psychosis[21,22]. In fact, vulnerable groups have a greater need to exploit the advantages of artificial intelligence technology for psychosis risk screening[12,14,80]. In reality, a "digital divide" and a "knowledge gap" still exist between urban and rural youths, the accessibility of digital technologies and services is unevenly distributed, and some youths still do not have the opportunity to access advanced technologies[81,82]. Furthermore, this unevenness is exacerbated by intergenerationally maintained inequality[46,78].

In addition to ethical issues, artificial intelligence-assisted psychosis risk screening in adolescents faces several other issues, including: (1) Small sample sizes: the use of machine learning to establish prediction models with high accuracy and strong generalization ability requires large samples[83], but

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many studies have mentioned the problem of too few samples, resulting in overfitting, which may lead to model errors and low accuracy[21,84,85]; (2) The prediction model must be optimized: for the same problem and the same sample set, the prediction accuracies of prediction models based on different algorithms vary, and the applicable scope and characteristics of each algorithm are different[86,87]; (3) Compliance: the level of the technical knowledge of patients is a key factor affecting their compliance [88], which also directly affects the accuracy of conclusions; and (4) Research bias: most research samples are active users of social media or patients who are informed in advance, while broader groups of real-world patients are not included; thus, the representativeness of the samples and generalizability of the findings may be limited.

LIMITATIONS

This study has two limitations. First, this paper is a narrative review and provides an outlook. This paper discusses the current status and challenges of using artificial intelligence-assisted methods for screening adolescents for psychiatric disorders, but we do not use an explicit approach, such as a systematic review. Therefore, the study is primarily intended to evoke research interest in the field but cannot be directly applied to clinical care. Second, this paper only summarizes the relevant literature in English and does not consider the literature published in languages other than English.

CONCLUSION

While focusing on the psychosis risk faced by adolescents worldwide, this paper reviews the influencing factors of adolescent psychosis risk, which can be divided into the following three main categories: national and social factors, family factors, and individual factors. This paper summarizes the benefits of artificial intelligence-assisted psychosis risk screening in adolescents, which are mainly manifested in improving the speed and timeliness of screening for those who are already sick and those with a potential risk of disease, promptly correcting the risky behavior of patients to prevent the occurrence and further aggravation of symptoms, and improving the accuracy and objectivity of screenings and the screening is discussed in detail. The advantage of chatbots in psychosis risk screening is that they can provide services to those with psychosis who have limited resources or accessibility problems, although privacy concerns and other ethical issues may exist. The accuracy of large-scale social media data analysis is gradually improving, and more technologies based on multimodal perception, understanding, and natural dialog and interactions are still needed to help comprehensively and accurately screen for psychosis risk in adolescents.

After surveying the current literature concerning artificial intelligence-assisted adolescent psychosis risk screening, we found that although artificial intelligence has been gradually applied to early psychosis risk screening, it has rarely been applied in studies that directly use adolescents as subjects. In view of the prevalence and harm of psychosis among adolescents worldwide, the timely screening of adolescent psychosis risks with artificial intelligence technology has considerable prospects for development. Furthermore, scientific progress must follow relevant ethical principles, not ignore vulnerable groups of adolescents, and ensure that artificial intelligence-assisted psychosis risk screening is conducted in an ethically acceptable manner, thereby minimizing potential adverse effects.

Based on the current status of psychiatric artificial intelligence research and practice, we propose that ethical issues constitute the main challenge of artificial intelligence-assisted psychosis risk screening in adolescents. The four biomedical ethics principles (respect for autonomy, nonmaleficence, beneficence, and justice) should be strictly obeyed. In addition to ethical issues, artificial intelligence-assisted psychosis risk screening in adolescents faces problems, such as small sample sizes, unoptimized prediction models, compliance, and research bias.

We propose that assuming compliance with ethical requirements, three main directions can be considered for artificial intelligence-assisted psychosis risk screening in adolescents in the future. First, we should develop nonperceptual real-time artificial intelligence screening with the help of technological advancements, such as 5G technology and the Internet of Things, to allow both the collection of individual emotional and health data and the prediction of individuals' mental health status in real time. Second, we should further reduce the cost of artificial intelligence-assisted screening. Psychosis is an important part of human health, and both poor and rich people should enjoy the benefits of technological progress. The long-term goal of artificial intelligence-assisted psychosis risk screening is that users should not pay high prices for the screening. Third, we should improve the ease of use of artificial intelligence-assisted screening techniques and tools such that regardless of an individual's level of knowledge, he or she can easily use artificial intelligence tools to screen for a psychosis risk.

FOOTNOTES

Author contributions: Liu XQ designed the study; Cao XJ and Liu XQ wrote the manuscript and managed the literature analyses; all authors approved the final manuscript.

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Observational Study

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ORIGINAL ARTICLE

Overlap of orthorexia, eating attitude and psychological distress in some Italian and Spanish university students

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Abstract

BACKGROUND

Orthorexia nervosa (ON) is the persistent concern of maintaining the self-imposed diet to improve one's health. Many factors have been associated to ON in university students.

AIM

To assess the prevalence of ON in Italian and Spanish university students in relation to eating attitude and psychological distress, and the possible overlaps between ON (evaluated with different scored questionnaires from the originally proposed ORTO-15), distress and risk of eating disorders.

METHODS

This study was carried out on 160 students recruited at La Sapienza University of Rome and at the Catholic University of Murcia. Questionnaires were administered to evaluate ON (ORTO-15 and sub-scores), body concerns (Multidimensional Body-Self Relations Questionnaire, MBSRQ, and Body Uneasiness test, BUT), psychological distress (Kessler Psychological Distress Scale, K10), physical activity (International Physical Activity Questionnaire, IPAQ), eating attitude (Eating Attitudes Test, EAT-26) and malnutrition (Starvation Symptom Inventory, SSI). Sex differences, within the same country, and differences between Italian and Spanish students, within the same sex, were evaluated.

RESULTS

The ORTO-15 positive subjects, assessed with the originally proposed cut-off, were above 70% in both Italian and Spanish students, with a higher prevalence in



the Spanish sample (Italian females 76.3%, Italian males 70.7%; Spanish females 97.0%, Spanish males 96.3%). According to ORTO-7, about 30% of Italian and 48% of Spanish students were positive to ON with no significant sex differences. When excluding students underweight (UW), overweight (OW) or obese (OB), as well as those potentially at risk of eating disorders or presenting mild, moderate and severe distress, in the resultant normal weight (NW)-K10^{neg}-EAT-26^{neg} subgroup, we did not find many correlations observed in the whole sample, including those between ORTO scores and BUT, SSI, Total MBSRQ and some of its components. Moreover, ORTO-7 resulted in the only ON score unrelated with Body Mass Index, MBSRQ components and IPAQassessed intense activity, in the NW-K10neg-EAT-26neg subgroup. After this sort of "exclusion diagnosis", the prevalence of ON of these students on the overall sample resulted in 16.9%, 12.2%, 15.2% and 25.9% for Italian females, Italian males, Spanish females and Spanish males, respectively.

CONCLUSION

In some university students ON could be a symptom of other conditions related to body image concerns and distress, as well as to high physical activity and appearance, fitness, health or illness orientation (from MBSRQ). However, ORTO-7 became independent from these confounding variables, after the exclusion of UW, OW, OB and students positive to EAT-26 and K10, suggesting the possibility of identifying orthorexic subjects with this specific questionnaire.

Key Words: Diet; Exercise; Food avoidance; Other Specified Feeding and Eating Disorder; Lifestyle

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Core Tip: This study explores the overlap of orthorexia nervosa with eating attitude and psychological distress in Italian and Spanish university students. After excluding, among normal weight students, those with high score on the Kessler Distress Scale and Eating Attitudes Test, we did not find correlations among orthorexia and Starvation Symptoms Inventory, Body Uneasiness Test, and Multidimensional Body Self-Relations Questionnaire (MBSRQ), observed in the whole sample. After this kind of "exclusion diagnosis", sub-scores of MBSRQ indicating body concerns correlated with ORTO-12 and ORTO-9, whereas ORTO-7 resulted the only score unrelated with all outcomes, including fitness and health orientations (MBSRQ), and intense physical activity.

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INTRODUCTION

The term orthorexia nervosa (ON) is referred to the psychological obsession with a healthy, organic and pure diet, and it is often based on stereotyped or erroneous nutritional beliefs, which can lead to dietary restrictions with resulting nutritional deficiencies [1-4]. Although ON shares several aspects in common with Anorexia Nervosa (AN)[5] and other eating disorders (EDs)[6], and around 70% of health professionals[7,8] believe that ON should be a distinct, clinically recognized ED, it is not included in the ICD-11 (International Statistical Classification of Diseases and Related Health Problems)[9,10]. In the Diagnostic and Statistical Manual of Mental Disorders, the terminology "eating disorders" has been changed to "Feeding and Eating Disorders", including eight categories: AN, bulimia nervosa, binge ED, pica, rumination, Avoidant or Restrictive Food Intake Disorder (ARFID), Other Specified Feeding and Eating Disorder (OSFED), and unspecified feeding and EDs[11]. Among these, ON has yet to be recognized as a separate ED, but has been included in OSFED, whereas ARFID refers to a persistent avoidance of food for various reasons (depressed temperament, phobias, sensory repulsion of the appearance, smell, texture of food), not necessarily associated with the search for a healthy diet that characterizes the subject with ON.

Although individuals with ON have an obsessive focus on healthy eating and may eliminate entire food groups, fearing they are unhealthy, they can later develop a typical ED[7,11]. Despite the awareness of clinicians on this condition [7,8,12], it is assumed that the estimate of subjects affected by ON is very complex, given the lack of explicit diagnostic criteria[13-16]. In particular, it has been



recently pointed out that the conflicting data on the prevalence of ON depend on differences in the tools used and in the cut-off points[17]. Moreover, authors concluded that the use of the ORTO-15 questionnaire to diagnose ON is questionable due to a high percentage of false-positive results[17], and different cut-offs or sub-scores have been proposed [18-24]. Abdullah et al [18] have suggested that a cutoff point of 35 is preferable than a cut-off at 40, and have reported that ON tendency is affected by sex and body mass index (BMI), and not affected by educational level. No differences were found in the prevalence of ON among students attending health-scientific, economic-humanistic, sport sciences and dietetics and nutrition university courses [25,26]. However, the reported prevalence of ON in university students was variable and affected by EDs, dieting and a high level of physical activity (PA)[25-30].

Among the criticisms relative to ORTO-15, Roncero et al^[31] highlighted the risk of including those who are on-diet among the individuals with ON, and the redundancy between the ORTO-15 and the Eating Attitudes Test (EAT-26). Accordingly, an overlap between ORTO-15 and EAT-26 has been reported in university students[32,33], but individuals with ON had lower psychological distress than those with ED risk[33]. Furthermore, it has been reported that overweight (OW) preoccupation and appearance orientation (AO), assessed with the Multidimensional Body-Self Relations Questionnaire-Appearance Scale (MBSRQ-AS), were significant predictors of ON[34].

The aim of this observational study was to evaluate the prevalence of ON in Italian and Spanish university students in relation to eating attitude and psychological distress, and the possible overlaps between ON (evaluated with different scored questionnaires), distress and risk of EDs. Moreover, a decision tree for the exclusion diagnosis of ON in healthy university students has been proposed.

MATERIALS AND METHODS

Study design, recruitment, and data collection

This study was carried out on a sub-group of participants from a previous study [35]. Undergraduate and doctoral students aged between 18 years and 35 years were recruited at La Sapienza University of Rome and at the Catholic University of Murcia. All the volunteers included in the study signed the informed consent, accompanied by an informative note, and the recruiter assigned them an alphanumeric code to guarantee privacy during the data management (all details about recruitment and protocol are available online at https://www.clinicaltrials.gov/ct2/show/NCT04099420). Body mass and stature were measured with the OMRON BF511 electronic scale and the SECA 217 portable stadiometer, respectively. BMI was calculated with the following formula: BMI = weight (kg)/height squared (m²). Moreover, several standardized questionnaires were administered.

ON

The ORTO-15 test is a questionnaire made up of 15 multiple-choice items based on a Likert scale (always, often, sometimes, never)[36]. The items concern three areas: the cognitive-rational area (items 1, 5, 6, 11, 12, 14), the clinical area (items 3, 7-9, 15), mainly related to anxiety-inducing and obsessive psychiatric disorders, and the emotional area (items 2, 4, 10, 13)[36]. A score of 1 was given to responses that were more indicative of ON, whereas a score of 4 was attributed to those that indicated a normal eating behavior. Therefore, lower scores correspond to a more pathological behavior, and total score ranges from 15 to 60, with the cut-off point equal to 40[36]. This value has been questioned as it was considered too high; consequently, a lower cut-off (35) was also chosen [18,19].

Additional versions of the ORTO test have been developed based on the selection of ORTO-15 items that more specifically could be indicative of the presence of symptoms of ON. ORTO-12 is a shorter version of 12 items, obtained by excluding items 5, 6 and 8 from the original ORTO-15, since they contribute less to the definition of ON, but a specific cut-off point was not determined[20]. ORTO-11 excludes items 5, 8, 14 and 15 of the ORTO-15. Final score ranges from 11 to 44, and a cut-off point < 25 has been considered the most appropriate to suggest a tendency to ON[21]. The ORTO-9 is a version of ORTO test which excludes items 1, 2 and 8, with a cut-off < 26.7. This test was found to be ineffective in predicting ON[22]. The ORTO-7 is based on items (1, 3, 4, 7, 9, 11 and 13) that mostly highlight the presence of ON with a cut-off $\leq 19[23,24]$.

Body image concerns

Two questionnaires were used to evaluate body image concerns, the MBSRQ[37,38] and the Body Uneasiness Test (BUT).

We evaluated the total MBSRQ and the MBSRQ Factor Subscales (FSs): AO, fitness orientation (FO), health orientation (HOr) and illness orientation (IO). In addition to these FSs, we evaluated the MBSRQ subscales: body areas satisfaction scale (BASS) and OW preoccupation (OP). Each of the MBSRQ scales has its corresponding items that can be answered by a primary number from 1 (strongly disagree) to 5 (strongly agree). The score of contraindicative items (6, 15-17, 23, 25, 28, 32-34, 36-38, 40, 42, 43, 45, 47-49) is reversed (*i.e.*, 1 = 5, 2 = 4, 4 = 2, 5 = 1). MBSRQ subscale scores are the means of the constituent items. The MBSRQ-AS is the shorter (34-item version) form of MBSRQ that assesses only the appearancerelated components of the body image construct^[37].



On the other hand, BUT includes two parts, BUT A (34 statements) and BUT B (37 body parts). Items are rated on a 6 points Likert-type scale (range 0-5, from "never" to "always") and high rates indicate greater body uneasiness^[39]. In addition to the total score, we evaluated BUT A weight phobia (WP, fear of being or becoming fat) and body image concerns (BIC, worries related to physical appearance) components.

Eating attitudes and malnutrition

Eating Attitudes Test (EAT-26) and Starvation Symptom Inventory (SSI) were administered to participants. The former is a standardized measure of symptoms and concerns characteristic of EDs, whereas the latter can reveal the presence of malnutrition.

EAT-26 is made up of 26 items and represent a screening tool to assess "ED risk" [40]. Although it does not provide a diagnosis of ED, the EAT-26 items include the subscales: "dieting" scale (items 1, 6, 7, 10, 11, 12, 14, 16, 17, 22, 23, 24, 26) and "bulimia and food preoccupation" scale (items 3, 4, 9, 18, 21, 25). Four behavioral questions are included to determine the presence of extreme weight-control behaviors. These items assess self-reported binge eating, self-induced vomiting, use of laxatives and treatment for ED over the preceding 6 mo. Participants were required to judge whether the item applied "always", "very often", "often", "sometimes", "rarely" or "never". Each extreme response in the "anorexic" direction is scored as a worth of 3 points, while the adjacent alternatives are weighted as 2 points and 1 point, respectively. A score \geq 20 on the EAT-26 does not necessarily mean that respondent has an ED. However, it indicates a high level of concern about dieting, body weight or problematic eating behaviors.

SSI is a 16-item questionnaire and participants were asked to estimate the number of days out of the preceding 28, in which they had experienced symptoms of starvation (hunger, poor concentration, heightened satiety, dizziness, reduction in rate of weight loss) on a 7-point Likert scale: never (0), 1-5 d (1), 6-12 d (2), 13-15 (3), 16-22 (4), 23-27 (5), and always (6)[41]. The highest score indicates increased frequency of starvation symptoms over the last 28 d.

Psychological distress and PA

The Kessler Psychological Distress Scale (K10)[42] and the International Physical Activity Questionnaire (IPAQ)[43] were used to evaluate distress and PA level, respectively.

K10 is a 10-item questionnaire about emotional states, each with a five-level response scale. The measure can be used as a brief screen to identify levels of distress[42]. Each item is scored from 1 (none of the time) to 5 (all of the time). Scores from the 10 items were then summed, yielding a minimum score of 10 and a maximum score of 50. According to the total score, the likelihood of having a mental disorder (psychological distress) is established^[42]; in particular, 10-19 likely to be well, 20-24 likely to have a mild disorder, 25-29 likely to have a moderate disorder, and 30-50 likely to have a severe disorder.

The IPAQ (short form)[43] includes items assessing the frequency and duration of PA in three ranges of intensity: intense PA (8.0 metabolic equivalent of tasks: METs), moderate PA (4.0 METs), and walking fast (3.3 METs), moderate (3.0 METs) and slow (2.5 METs) pace[43]. Based on collected data about the frequency and duration of PA, energy expenditure (expressed as MET-min/wk) has been estimated. One MET is the rate of energy expenditure at rest, and it is approximately equal to 3.5 mL O2 kg-1 min-1 in adults. According to the Italian Society of Endocrinology, IPAQ allows to classify population in three PA levels: Low (the lowest level of PA, less than 700 METs-min/wk), Moderate (Total PA between 700 and 2519 METs-min/wk) and High (Total PA of at least 2520 METs-min/wk).

Statistical analysis

Categorical variables were expressed as percentages and significance assessed by the χ^2 test. Continuous variables showing a normal pattern (normality test Shapiro-Wilk passed) were expressed as means with SD, otherwise data were expressed as median (25%-75% range). Results were analyzed by analysis of variance (ANOVA, Shapiro-Wilk test passed), or by Kruskal-Wallis one-way analysis of variance on ranks (Shapiro-Wilk test failed). The significance of the differences between females and males within the same country, and those between the different countries within the same sex, were evaluated using the Student-Newman-Keuls method (Shapiro-Wilk test passed) or the Dunn's method (Shapiro-Wilk Test failed). Spearman correlation was performed between variables. The level of significance was set below 5% (*P* < 0.05).

RESULTS

Country and sex differences

Characteristics of students and differences between Italy and Spain (evaluated within the same sex), and sex differences (evaluated within the same country) are reported in Table 1. Only in Italy, females (IT-F) had a significant lower BMI than males (IT-M), whereas the different prevalence of underweight (UW),



Table 1 Students' characteristics					
	IT-F, <i>n</i> = 59	IT-M, <i>n</i> = 41	SP-F, <i>n</i> = 33	SP-M, <i>n</i> = 27	
Age (yr)	24 (23-28)	25 (23-28)	23 (21-28)	24 (24-26)	
Height (m)	1.64 ± 0.06^{a}	1.78 ± 0.08^{a}	1.63 ± 0.06^{a}	1.78 ± 0.07^{a}	
Weight (kg)	55.9 (53.0-61.0) ^a	77.0 (70.9-87.5) ^a	58.7 (52.4-66.5) ^a	76.3 (67.1-82.0) ^a	
BMI (kg/m ²)	20.7 (19.4-22.4) ^a	24.5 (22.1-27.1) ^a	23.3 (20.2-24.5)	24.1 (21.8-25.5)	
Underweight (%)	11.9	0.0	9.1	3.7	
Overweight (%)	6.8	31.7	15.1	25.9	
Obese (%)	5.1	9.7	3.0	3.7	
IPAQ (MET-min/wk)	2232 (1080-4986)	4380 (2305-6277)	2100 (1202-4395) ^a	4970 (2575-6780) ^a	
IPAQ walking	630 (315-1260)	700 (488-1323)	525 (244-1230)	600 (240-1470)	
IPAQ moderate	720 (160-1680)	600 (240-1440)	600 (240-1440)	720 (120-1440)	
IPAQ intense	480 (0-1920) ^a	2160 (400-3840) ^a	800 (0-1680) ^a	2400 (960-4320) ^a	
PA low (%)	16.9	7.3	18.2	7.4	
PA moderate (%)	40.7	19.5	36.4	14.8	
PA high (%)	42.4	73.2	45.5	77.8	
Smokers (%)					
Habitual	15.3	19.5	9.1	7.7	
Occasional	16.9	19.5	15.2	15.4	
Ex-smokers	8.5	9.8	9.1	3.8	
K10	15.0 (13.0-19.0)	14.0 (12.0-17.0)	19.0 (15.0-27.5)	17.0 (14.0-20.0)	
Mild (%)	11.9	0.0 ^b	15.2	25.9 ^b	
Moderate (%)	6.9	9.8	18.2	3.7	
Severe (%)	5.1	4.9	15.2	3.7	

 $^{a}P < 0.05$ between females and males within country.

 $^{b}P < 0.05$ between Italians and Spaniards within sex.

Categorical variables are expressed as percentages. Continuous variables are expressed as means with standard deviation (Shapiro-Wilk Test passed), or as median (25%-75% range, Shapiro-Wilk Test failed). BMI: Body mass index; IPAQ: International Physical Activity Questionnaire; IT-F: Italian females; IT-M: Italian males; K10: Kessler Psychological Distress Scale; SP-F: Spanish females; SP-M: Spanish males; PA: Physical activity.

> OW and obese (OB) students, as well as of volunteers who practice Low, Moderate or High PA, did not reach significance in both countries (Table 1). However, Spanish males (SP-M) on average practiced more PA than females (SP-F), and sex differences were found in IPAQ intense (MET-min/wk from intense activities) in both countries (Table 1). No differences were found in smoking habits, whereas a high percentage of SP-M reported mild psychological distress (assessed by K10) than IT-M (Table 1). SP-M were also those with higher total MBSRQ, MBSRQ-AO, -FO, -HOr, -AS and -BASS, whereas no differences were found in MBSRQ-IO and -OP (Table 2). On the contrary, IT-F had higher BUT-A - BIC and - WP than IT-M, but lower compared to SP-F (Table 2). Differences in total BUT and in its components (BUT-A and BUT-B) did not reach statistical significance (Table 2).

> Concerning ORTO-15, lower values (indicating high ON) were observed in Spanish students compared to Italians, when using the first proposed cut-off point of 40[36], and with ORTO-12, -11 and -9 (Table 3). By using the lower cut-off point of 35 for ORTO-15 or the ORTO-11, -9, and -7, IT-F resulted with less tracts of ON than SP-F (Table 3). Both SP-F and SP-M had higher EAT-26 total score than Italian counterparts, despite dieting and bulimia components did not reach statistical significance (Table 3). Similar results came from SSI, suggesting more starvation symptoms in Spaniards compared to Italians (Table 3).

Overlaps and correlations among outcomes

Figure 1A illustrates the prevalence of ON in the whole sample (n = 160), by using different cut-off points and scores for ORTO questionnaire, among students who presented ED risk (EAT-26) or mild, moderate, and severe psychological distress (K10). In addition to the overlaps among these conditions

Table 2 Body image concerns				
	IT-F, <i>n</i> = 59	IT-M, <i>n</i> = 41	SP-F, <i>n</i> = 33	SP-M, <i>n</i> = 27
MBSRQ	227.0 ± 21.7	231.6 ± 27.6 ^b	227.9 ± 24.3^{a}	249.3 ± 27.5 ^{a,b}
MBSRQ-AO	3.3 ± 0.9	3.2 ± 0.5^{b}	3.5 ± 0.5	3.5 ± 0.5^{b}
MBSRQ-FO	3.4 (3.0-3.9)	3.5 (3.2-4.2)	3.5 (3.0-4.0) ^a	4.2 (3.6-4.4) ^a
MBSRQ-HOr	3.4 ± 0.5	3.4 ± 0.6	3.2 ± 0.6^{a}	3.6 ± 0.6^{a}
MBSRQ-IO	3.3 ± 0.6	3.3 ± 0.5	3.4 ± 0.5	3.6 ± 0.6
MBSRQ-AS	3.1 ± 0.3	3.1 ± 0.4^{b}	3.1 ± 0.4^{a}	$3.4 \pm 0.4^{a,b}$
MBSRQ-BASS	3.2 ± 0.7	3.4 ± 0.6	3.1 ± 0.8^{a}	3.5 ± 0.7^{a}
MBSRQ-OP	2.0 (1.8-2.8)	2.5 (1.8-3.0)	2.5 (2.1-3.0)	2.5 (2.0-3.3)
BUT	33.0 (19.0-61.0)	16.0 (6.0-39.0)	60.0 (30.5-82.0)	35.0 (21.0-74.0)
BUT-A	16.0 (9.0-36.0)	10.0 (2.0-20.0)	33.0 (12.5-51.0)	18.0 (13.0-42.0)
BUT-A - WP	0.9 (0.4-1.6) ^b	0.5 (0.1-1.0)	1.8 (0.8-2.6) ^b	1.1 (0.6-1.8)
BUT-A - BIC	0.7 (0.2-1.3) ^a	0.2 (0.0-0.9) ^a	0.9 (0.5-1.6)	0.6 (0.2-1.6)
BUT-B	15.0 (7.0-31.0)	11.0 (3.5-21.0)	25.5 (14.0-42.5)	21.0 (8.0-36.0)

 $^{a}P < 0.05$ between females and males within country.

 $^{b}P < 0.05$ between Italians and Spaniards within sex.

Categorical variables are expressed as percentages. Continuous variables are expressed as means with standard deviation (Shapiro-Wilk Test passed), or as median (25%-75% range, Shapiro-Wilk Test failed). AO: Appearance orientation; AS: Appearance scales; BASS: Body areas satisfaction scale; BIC: Body image concerns; BUT; Body Uneasiness Test; FO; Fitness orientation; HOr; Health orientation; IO; Illness orientation; IT-F; Italian females; IT-M; Italian males; MBSRQ: Multidimensional Body-Self Relations Questionnaire; OP: Overweight preoccupation; SP-F: Spanish females; SP-M: Spanish males; WP: Weight phobia.

> and ON, psychological distress was observed both in students presenting ED risk or in those resulting negative to the EAT-26 test, without sex differences (Figure 1B). On the other hand, in Figure 2 is presented the prevalence of ON among different BMI classes and lifestyle factors, such as PA level and smoking habits.

> From the aforementioned results, and in order to reduce the potential confounder as being on caloric restriction[31], Spearman correlations were evaluated in both the total sample (n = 160, 92 F and 68 M) and a subgroup of students (n = 66, 38 F and 28 M) with normal weight (NW), excluding those who suffered from mild, moderate and severe distress, or potentially at ED risk (NW-K10^{neg}-EAT-26^{neg}).

> As regards the NW-K10^{neg}-EAT-26^{neg} group, all subjects resulted with ON when ORTO-15 (40 cut-off point) was applied, whereas a prevalence of 37.9% was observed applying ORTO-15 (35 cut-off point), 18.2% considering ORTO-11, 22.7% with ORTO-9, and 40.0% with ORTO-7. Concerning lifestyle, the percentage of non-smokers (60.6%) was higher (P < 0.05) compared to those of ex-smokers (7.6%) and smokers (13.6%) among students with ON, assessed with ORTO-15 (40 cut-off). Similar results were reported when ORTO-15 (35 cut-off point) was used, with a prevalence of ON in non-smokers and smokers corresponding to 60.0% and 16.0%, respectively, as well as for results of ORTO-11 (prevalence of ON: non-smokers 66.7% and smokers 16.7%), ORTO-9 (prevalence of ON: non-smokers 66.7% and smokers 13.3%) and ORTO-7 (prevalence of ON: non-smokers 74.1% and smokers 7.4%). Among the students with ON, the percentage of those practicing high PA was higher (P < 0.05) compared to low PA for ORTO-15 (cut-off 40: high PA 54.4% and low PA 15.2%; cut-off 35: high PA 68.0% and low PA 8.0%), ORTO-11 (high PA 75.0% and low PA 0%), ORTO-9 (high PA 80.0% and low PA 0.0%) and ORTO-7 (high PA 59.3% and low PA 11.1%).

> Considering the whole sample, among the different ORTO scores a relationship between ORTO-9 and BMI was found, and Table 4 depicts the Spearman correlations between ORTO scores and those from the other questionnaires. All ORTO scores were inversely correlated to SSI (Table 4), which indicates that high ON corresponds to more starvation symptoms since ORTO has reverse scores. Similarly, all ORTO scores were correlated to EAT-26 and its components (Table 4). On the contrary, the relationships between each ORTO score and BUT components were different (Table 4). Weight phobia (BUT A-WP) was correlated with ORTO-12, ORTO-9 and ORTO-7. The latter resulted the only one not related to MBRSQ-AS, MBSRQ and its health component (MBSRQ-HOr), as well as to the MBSRQ-OP (Table 4). Moreover, the correlation coefficients between ORTO scores and MBSRQ-AO and -FO were lower for ORTO-7 compared to the other scores (Table 4). On the other hand, ORTO-7 resulted in the only one that was related to the MBSRQ-BASS, and ORTO-9 was the only score related to IPAQ and to its intense activities and walking components (Table 4).



Table 3 Orthorexia, eating attitude and malnutrition				
	IT-F, <i>n</i> = 59	IT-M, <i>n</i> = 41	SP-F, <i>n</i> = 33	SP-M, <i>n</i> = 27
ORTO-15	36.8 ± 3.4^{a}	36.4 ± 34.2^{a}	34.2 ± 3.6^{a}	33.8 ± 3.4^{a}
Cut-off 40 (%)	76.3 ^a	70.7 ^a	97.0 ^a	96.3 ^a
Cut-off 35 (%)	23.7 ^a	36.5	48.5 ^a	55.5
ORTO-12	30.7 ± 3.0^{a}	30.4 ± 3.4^{a}	27.5 ± 2.9^{a}	27.2 ± 2.7 ^a
ORTO-11	27.8 ± 2.8^{a}	27.6 ± 3.4^{a}	25.2 ± 2.9^{a}	25.4 ± 3.0^{a}
Cut-off 25 (%)	11.9 ^a	19.5 ^a	36.4 ^a	25.9 ^a
ORTO-9	30.4 ± 3.4^{a}	30.0 ± 3.7^{a}	27.8 ± 3.2^{a}	27.1 ± 3.2 ^a
Cut-off 26.7 (%)	13.6 ^a	19.5 ^a	30.3 ^a	40.7 ^a
ORTO-7	20.0 (18.0-22.0) ^a	19.0 (18.0-21.0)	19.0 (15.5-19.5) ^a	19.0 (16.0-19.0)
Cut-off 19 (%)	32.2	29.3	48.5	48.1
EAT-26	6.0 (3.0-10.0) ^a	5.0 (3.0-8.5) ^a	11.0 (5.0-17.5) ^a	13.0 (7.0-17.0) ^a
Cut-off 20 (%)	8.5	4.9	21.2	11.1
Dieting	2.0 (0.0-5.0)	2.0 (0.04.5)	4.0 (0.5-9.5)	6.0 (3.0-9.0)
Bulimia	3.0 (3.0-3.0)	3.0 (3.0-3.0)	3.0 (3.0-5.5)	3.0 (3.0-5.0)
SSI	14.0 (8.0-22.0) ^a	10.0 (4.5-16.0) ^a	24.0 (20.0-43.0) ^a	18.0 (12.0-30.0) ^a

 $^{a}P < 0.05$ between Italians and Spaniards within sex. Not significant between females and males within country.

Categorical variables are expressed as percentages. Continuous variables are expressed as means with standard deviation (Shapiro-Wilk Test passed), or as median (25%-75% range, Shapiro-Wilk Test failed). EAT-26: Eating Attitudes Test; IT-F: Italian females; IT-M: Italian males; ORTO: Scores for orthorexia nervosa (ON, reverse scores, lower values indicate high ON); SP-F: Spanish females; SP-M: Spanish males; SSI: Starvation Symptom Inventory.

In the NW-K10^{neg}-EAT-26^{neg} subgroup, we did not find many correlations as observed in the whole sample, including those between ORTO scores and BUT, SSI, total MBSRQ and some of its components (Table 4). Although no relationship was found among ORTO score and SSI in the NW-K10^{neg}-EAT-26^{neg} subgroup, SSI was correlated with MBSRQ-OP (0.291, P < 0.05). The latter, as well as MBSRQ-AO and - AS, was correlated with ORTO-12 and ORTO-9 (Table 4), but also with BMI (0.260, P < 0.05).

On the other hand, in the NW-K10^{neg}-EAT-26^{neg} subgroup, MET-min/wk from intense activities correlated with ORTO-15, -12, -11 and -9 (Table 4), as well as with BMI (0.442, P < 0.001), MBSRQ-FO (0.629, P < 0.001), -HOr (0.387, P < 0.01), that were highly related (coefficient of correlation MBSRQ-FO (0.629, P < 0.001) and, to a lesser extent, IPAQ-intense activity was related to MBSRQ-OP (0.253, P < 0.05). MBSRQ-IO correlated with both MBSRQ-HOr (0.447, P < 0.001) and -AO (0.256, P < 0.05), that was related to MBSRQ-FO (0.329, P < 0.01). Interestingly, ORTO-7 resulted in the only score unrelated neither with BMI nor with the other evaluated outcomes in the NW-K10^{neg}-EAT-26^{neg} subgroup (Table 4). The prevalence of ON from ORTO-7 in students included in the NW-K10^{neg}-EAT-26^{neg} subgroup on the overall sample resulted to be 16.9%, 12.2%, 15.2% and 25.9% for IT-F, IT-M, SP-F and SP-M, respectively.

DISCUSSION

In light of the reported overlaps of ON with other conditions and of the criticism highlighted from literature about ORTO-15[17-34], we have evaluated the prevalence of ON among 160 Italian and Spanish university students, from the 194 recruited in a previous study[35], who agreed to fill the standardized questionnaires: EAT-26, K10, BUT, MBSRQ, and IPAQ. The sample had, on average, a medium adherence to the Mediterranean diet and a low risk of excessive alcohol consumption[35]. Concerning country differences, the prevalence of mild distress was higher in SP-M compared to IT-M (Table 1), whereas IT-F had lower BUT-A -WP than SP-F (Table 2). SP-M had higher total MBSRQ, MBSRQ-AO, -FO, -HOr, -BASS and -AS, whereas no differences were found in MBSRQ-IO and -OP (Table 2).

Higher SSI, EAT-26 and ON, by using ORTO-15 (40 cut-off), -12, -11 and -9, were observed in Spanish students compared to Italians, regardless of sex (Table 3). Accordingly, no significant difference between female and male students in ORTO-15 has been reported in university students[25,44,45].

Table 4 Spearman correlations					
	ORTO-15	ORTO-12	ORTO-11	ORTO-9	ORTO-7
BMI (<i>n</i> = 160)	NS	NS	NS	-0.184 ^a	NS
BMI NW-K 10^{neg} -EAT- 26^{neg} ($n = 66$)	-0.315 ^a	-0.304 ^a	-0.284 ^a	-0.432 ^c	NS
SSI (<i>n</i> = 160)	-0.169 ^a	-0.264 ^c	-0.223 ^b	-0.244 ^b	-0.228 ^b
EAT-26 (<i>n</i> = 160)	-0.363 ^c	-0.414 ^c	-0.349 ^c	-0.432 ^c	-0.357 ^c
Dieting $(n = 160)$	-0.429 ^c	-0.444 ^c	-0.387 ^c	-0.516 ^c	-0.307 ^c
Bulimia (<i>n</i> = 160)	-0.188 ^a	-0.263 ^c	-0.180 ^a	-0.304 ^c	-0.223 ^b
BUT (<i>n</i> = 160)	NS	NS	NS	NS	-0.160 ^a
BUT A – WP (<i>n</i> = 160)	NS	-0.187 ^a	NS	-0.181 ^a	-0.166 ^a
BUT B (<i>n</i> = 160)	NS	NS	NS	NS	-0.165 ^a
MBSRQ ($n = 160$)	-0.354 ^c	-0.279 ^c	-0.341 ^c	-0.414 ^c	NS
MBSRQ-AO (<i>n</i> = 160)	-0.333°	-0.296 ^c	-0.326 ^c	-0.359 ^c	-0.177 ^a
MBSRQ-AO NW-K 10^{neg} -EAT- 26 ^{neg} ($n = 66$)	NS	-0.285 ^a	NS	-0.289 ^a	NS
MBSRQ-FO (<i>n</i> = 160)	-0.361 ^c	-0.310 ^c	-0.358 ^c	-0.414 ^c	-0.180 ^a
MBSRQ-FO NW-K 10^{neg} -EAT- 26^{neg} ($n = 66$)	-0.304 ^a	-0.371 ^a	-0.373 ^a	-0.450 ^c	NS
MBSRQ-HOr (<i>n</i> = 160)	-0.341 ^c	-0.248 ^b	-0.333°	-0.402 ^c	NS
MBSRQ-HOr NW-K10 ^{neg} - EAT-26 ^{neg} $(n = 66)$	-0.358 ^b	-0.408 ^c	-0.433 ^c	-0.491 ^c	NS
MBSRQ-IO (<i>n</i> = 160)	-0.162 ^a	NS	-0.181 ^a	NS	NS
MBSRQ-BASS ($n = 160$)	NS	NS	NS	NS	-0.169 ^a
MBSRQ-OP (<i>n</i> = 160)	-0.177 ^a	-0.216 ^b	-0.165 ^a	-0.302 ^c	NS
MBSRQ-OP NW-K 10^{neg} -EAT- 26^{neg} ($n = 66$)	NS	-0.366 ^b	NS	-0.357 ^b	NS
MBRSQ-AS ($n = 160$)	-0.195 ^a	-0.155 ^a	-0.161 ^a	-0.265 ^c	NS
MBSRQ-AS NW-K 10^{neg} -EAT- 26^{neg} ($n = 66$)	NS	-0.244 ^a	NS	-0.306 ^a	NS
IPAQ ($n = 160$)	NS	NS	NS	-0.183 ^a	NS
Walking (<i>n</i> = 160)	NS	NS	NS	-0.162 ^a	NS
Intense activity ($n = 160$)	NS	NS	NS	-0.178 ^a	NS
Intense activity NW-K 10^{neg} -EAT- 26^{neg} ($n = 66$)	-0.404 ^c	-0.330 ^b	-0.273 ^a	-0.461 ^c	NS

 $^{a}P < 0.05.$

 $^{b}P < 0.01.$

 $^{c}P < 0.001.$

AO: Appearance orientation; AS: Appearance scales; BASS: Body areas satisfaction scale; BMI: Body mass index; BUT: Body Uneasiness Test; EAT-26: Eating Attitudes Test; FO: Fitness orientation; HOr: Health orientation; IO: Illness orientation; IPAQ: International Physical Activity Questionnaire; MBSRQ: Multidimensional Body-Self Relations Questionnaire; NS: Not significant; NW- K10^{neg}-EAT-26^{neg}: Normal weight students excluded those with K10 or EAT-26 positive test; OP: Overweight preoccupation; ORTO: Scores for orthorexia nervosa; SSI: Starvation Symptom Inventory; WP: Weight phobia.

> Among NW Polish university students, ORTO-15 correlated with MBSRQ-OP, -AO, -FO, -HOr and -BASS in females, whereas in male students, body image concerns were not associated with ON[38]. The Spearman correlations (Table 4) confirmed the previously reported relationship between ON, depending on ORTO score applied to the whole sample, and MBSRQ-OP, -AO[34], and EAT-26[32,33], and we have observed overlaps between ORTO scores, EAT-26 and K10 (Figure 1). Although the prevalence of UW, OW and OB students did not reach significance in the whole sample, IT-F had a lower BMI (Table 1) and higher BUT-A – BIC (Table 2) than IT-M.

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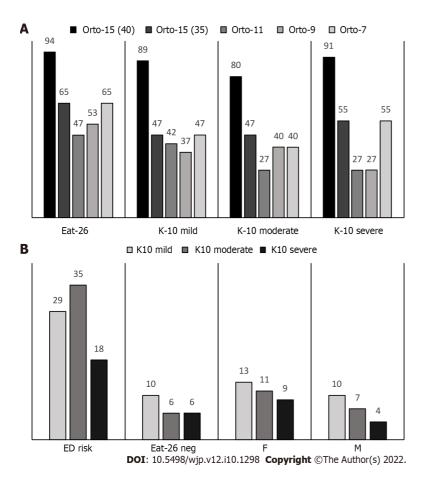


Figure 1 Prevalence of orthorexia nervosa. A and B: Prevalence of orthorexia nervosa among students who presented an eating disorder (ED) risk [Eating Attitudes Test (EAT)-26] or psychological distress [Kessler Psychological Distress Scale (K10)] (A) and overlap of EAT-26 and K10 (B), in the whole sample. ORTO: Scores for orthorexia nervosa; K10: Kessler Psychological Distress Scale; EAT-26: Eating Attitudes Test; F: Italian and Spanish females; M: Italian and Spanish males; ED: Eating disorder.

In order to reduce the potential confounder as being on a diet[31], we have evaluated a subgroup of NW students, excluding volunteers who presented distress or ED risk (NW-K10^{neg}-EAT-26^{neg}). All students in the NW-K10^{neg}-EAT-26^{neg} group had ON, when ORTO-15 with the 40 cut-off was applied, whereas the percentage of ON varied with ORTO-15 (35 cut-off), -11, -9 and -7. Concerning lifestyle, the percentage of non-smokers was higher compared to those of smokers among students with ON in the NW-K10^{neg}-EAT-26^{neg} group. In this context, it has been proposed to distinguish between ON and healthy orthorexia (HO), a non-pathological tendency to follow a healthy diet[46]. HO can be the successful result of dissemination campaigns aimed to increase nutrition knowledge from WHO and National recommendations, but the proposed etiology of ON includes high level of education, pseudoscientific nutritional news on social media and psychological factors[12]. From a study carried out in nutrition and dietetics, university students emerged that Instagram use might be considered as an ONrisk factor^[47]. Besides, K10 median of the scores of students enrolled in health-related study courses was higher than those of non-health-related degree courses[48]. Furthermore, it has been suggested that obsessive healthy eating fixations may increase the risk for ED in athletes and that more education and awareness are warranted to minimize the risk for ON and ED in student-athletes[30]. High level of PA in association with ON, assessed with ORTO-15 using cut-off scores of 35, was more often seen in men from sports science and less often in women from business course^[27]. With a cut-off of 40, ORTO-15 resulted lower among students who performed more than ten h/wk of exercise, regardless of the engagement in university sport teams, including athletes competing in aesthetic and weight dependent sports[28]. In our sample, sex differences were found in MET-min/wk from intense activities in both countries (Table 1). In the NW-K10^{neg}-EAT-26^{neg} group, among the students with ON, the percentage of those practicing high PA was higher compared to those having low PA, whereas we did not find correlations observed in the whole sample, including those between ORTO scores and BUT, SSI, and total MBSRO.

The present study has both strengths and limitations, taking into account the suggestion of a pilot study[49] that reported high levels of disparity among psychometric scores, including ORTO-15, EAT-26 and MBSRQ, recommending the use of multiple psychometric instruments for ON diagnosis. Furthermore, the evaluation of dietary intakes of 10 individuals (assessed using 24-h recall) failed to meet the guidelines for several nutrients[49]. As a point of strength, we have used different stan-



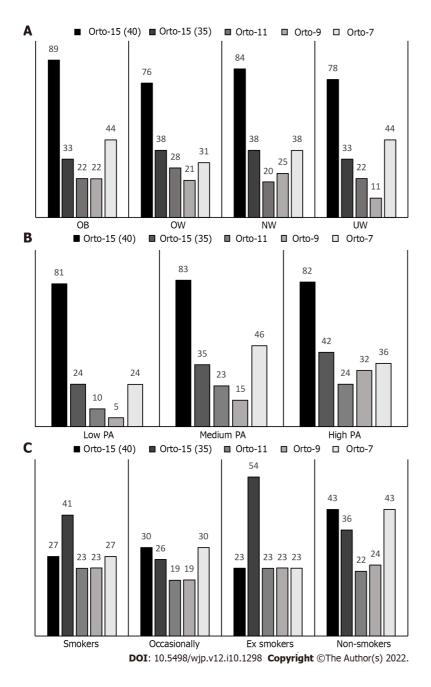
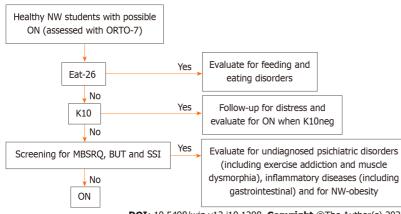


Figure 2 Prevalence of orthorexia nervosa among different body mass index classes and lifestyle factors. A: Body mass index; B: Physical activity; C: Lifestyle factors. ORTO: Scores for orthorexia nervosa; OB: Obese; OW: Overweight; NW: Normal weight; UW: Underweight; PA: Physical activity.

> dardized questionnaires in order to exclude students presenting overlaps. As a limitation, we did not evaluate the nutritional status. However, the relationships between SSI and all ORTO scores observed in the whole sample, but not in the NW-K10^{neg}-EAT-26^{neg} subgroup, suggest that students with more starvation symptoms were excluded from the applied exclusion criteria. In fact, the greatest strength is that ORTO-7 resulted in the only score unrelated neither with BMI nor with the other evaluated outcomes in the NW-K10^{reg}-EAT-26^{reg} subgroup (Table 4). The prevalence of ON from ORTO-7 after the "exclusion diagnosis" in the NW-K10^{neg}-EAT-26^{neg} subgroup on the overall sample ranged between 12.2% and 25.9%, values lower than those reported in the whole sample. However, it was higher than some observed with other sub-scores of ORTO test (Table 4). This finding suggests that, among the limitations, we did not include the OCD in the "exclusion diagnosis" [50]. However, Łucka *et al* [29] using ORTO-15 (score of 35 was considered as cut-off point), EAT-26 and Maudsley Obsessive Compulsive Inventory (MOCI), found that individuals with suspected ON (ORTO-15, score of 35) had higher BMI and EAT-26 score, whereas MOCI did not differ from ORTO-15 negative group. From that, authors suggested that ON meets the criteria of ED and not of OCD[29]. On the other hand, considering the relationship among other ORTO scores and body concerns, and high PA component of IPAQ, also questionnaires evaluating exercise addiction [51,52] and muscle dysmorphia [26] should be included in the "exclusion diagnosis".



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Figure 3 Decision tree for the exclusion diagnosis of orthorexia nervosa (ON) in healthy normal weight university students. ORTO-7: Score for orthorexia nervosa; K10: Kessler Psychological Distress Scale; EAT-26: Eating Attitudes Test; MBSRQ: Multidimensional Body-Self Relations Questionnaire; BUT: Body Uneasiness Test; SSI: Starvation Symptom Inventory; NW: Normal weight.

Among exclusion diagnosis (gastrointestinal disorders) there are functional dyspepsia [53], non-celiac gluten sensitivity (NCGS)[54,55] and irritable bowel syndrome (IBS)[55]. Both NCGS and IBS are more prevalent in young women compared to men[55], and up to 90% of patients with IBS exclude certain foods to improve their gastrointestinal symptoms [56]. Besides, negative effects of prescribing restrictive diets can be observed due to the association between ED and gastrointestinal symptoms[56]. In this context, the differential diagnosis and assessment of AN already began with the exclusion of diseases, including inflammatory bowel disease (Crohn's disease or ulcerative colitis), malignancies, thyrotoxicosis, diabetes, cerebral tumor, major depressive or schizophrenic illnesses[10].

Although the scientific community is divided into those who consider ON as a separate ED and those who do not[7], agreement exists on the needs of prevention (primary and secondary) and some diagnostic criteria, as reviewed by Atzeni et al[16], including: obsessive concern for healthy eating, fear anxiety and avoidance of certain foods components (additives, preservatives, fats or other elements considered unhealthy). Furthermore, there is broad consensus on the induction of malnutrition by ON and impacts on social and professional functioning. Other suggested criteria (not endorsed by all experts) included differences between ON and OCD or from schizophrenia, excessive time spent or rituals in preparing meals, excessive spending money for buying healthy foods; anxieties and fears concerning transgressions, and the exclusion of individuals who observe a religious practice or have medical problems^[16]. Our work suggests including EDs and psychological distress among the medical problems for the exclusion diagnosis of ON in healthy NW individuals. Otherwise, we suggest considering ON as a symptom of other diseases or a disease-induced comorbidity.

CONCLUSION

Although a generalization to the whole population should not be made, considering the recent suggestions on the need for further investigation of the comorbidity between ON and OCD across different cultural groups[50], it emerged that ON could be an indicator/symptom of other problems related to body image perception, as well as high PA, psychological distress, appearance, fitness, health, or IO, in some university students. Accordingly, recent studies have found relationships between ON, vigorous-intensity PA and dieting[47,57]. In our study, the ORTO-7 was found to be independent from these confounders, after the exclusion of UW, OW, OB, and EAT-26 and K10 positive students, suggesting the possibility of defining subjects with ON. Therefore, considering the overlap conditions, we suggest a decision tree for differential/exclusion diagnosis of ON (Figure 3). In order to identify the real orthorexic subjects among healthy students with NW, firstly the presence of EDs should be assessed, followed by the evaluation of the distress level, and lastly the presence of body image concerns and malnutrition (Figure 3). Moreover, a high percentage of students (25.5% males and 40.1% females) with NW obesity (NWO) have been reported, and stress management behavior decreased the risk of NWO in females [58]. In this context, Villa et al [47] observed that ON was associated not only with heavy exercise but also with sedentary behavior. In students with NWO, low PA could be associated with dieting, inducing ON. In conclusion, due to the several confounders and overlap conditions, flowcharts, diagnostic algorithms and a decision tree for differential diagnosis and management of ON should be included, as well as guidelines and consensus statements of experts in the future.

ARTICLE HIGHLIGHTS

Research background

Many factors have been associated to orthorexia nervosa in university students.

Research motivation

To assess the prevalence of orthorexia nervosa in Italian and Spanish university students.

Research objectives

To assess the prevalence of orthorexia nervosa in relation to eating attitude and psychological distress.

Research methods

Questionnaires were administered to evaluate orthorexia nervosa, body concerns, psychological distress, physical activity, eating attitude and starvation symptoms.

Research results

When excluding students underweight (UW), overweight (OW) or obese (OB), as well as those potentially at risk of eating disorders or presenting distress, in the resultant normal weight (NW)-K10neg. EAT-26^{neg} subgroup, we did not find many correlations observed in the whole sample, including those between ORTO scores and Body Uneasiness Test, Starvation Symptom Inventory, Total Multidimensional Body-Self Relations Questionnaire (MBSRQ) and some of its components. Moreover, ORTO-7 resulted the only ON score unrelated with Body Mass Index, MBSRQ components and IPAQ-assessed intense activity, in the NW-K10^{neg}-EAT-26^{neg} subgroup. After this sort of "exclusion diagnosis", ORTO-7 became independent from these confounding, after the exclusion of UW, OW, OB and students positive to EAT-26 and K10, suggesting the possibility of identifying orthorexic subjects with this specific questionnaire.

Research conclusions

In some university students ON could be a symptom of other conditions related to body image concerns and distress, as well as to high physical activity and appearance, fitness, health or illness orientation. ORTO-7 became independent from these confounding factors, after the exclusion of UW, OW, OB and students positive to EAT-26 and K10, suggesting the possibility of identifying orthorexic subjects with this specific questionnaire.

Research perspectives

Considering the overlap conditions, we suggest a decision tree for differential/exclusion diagnosis of ON.

FOOTNOTES

Author contributions: Peluso I and Villaño D contributed to the conceptualization; Aiello P, Toti E and Raguzzini A contributed to the investigation; Aiello P and Peluso I contributed to the original draft preparation; Toti E, Raguzzini A and Villaño D contributed to the review and editing; Peluso I contributed to the supervision.

Institutional review board statement: The study was conducted according to the guidelines of the Declaration of Helsinki, and approved by the Ethics Committee for Human Experimentation of the La Sapienza University of Rome (protocol code 1382/2019, approved on 16 July 2019) and by the Ethics Committee of the Catholic University of Murcia (UCAM) (protocol code CE071906, approved on 3 July 2019).

Informed consent statement: Informed consent was obtained from all subjects involved in the study. Volunteers did not sign consent to share single individual data, but only cumulative results.

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