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# A multi-attribute utility decision support tool for a smart campus—UAE as a case study

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The advancement in technologies in the education sector has improved living standards and acts as a sustaining factor for future development. Recently, the integration of technologies into the campus to transform it into a Smart Campus has experienced exponential growth in interest from researchers. Though various definitions of the concept of 'Smart Campus' have been proposed, the integration of the end users' perception is always lacking in the concept. This study, therefore, intends to build on the theory to classify the most significant criteria that underpin the 'Smart Campus' by considering the institute's stakeholders' perceptions. A multi-step methodological approach is adopted to develop a decision support tool that allows the decision makers to invest in the optimum solution to transform a traditional campus into a smart campus. The study initially looks into the criteria and sub-criteria from the literature that defines a 'Smart Campus'. Secondly, a survey was conducted by targeting a sample of students, faculty, administrative staff, and IT support personnel from a leading institute in the UAE region as a single case study. Thirdly, an AHP analysis was performed among different stakeholders. The findings suggested that there exists a consensus among the perception of a diverse group of participants who perceive smart security and safety, campus navigation, and adaptive learning as the most important criteria and applications to transform the traditional campus into a smart campus. Finally, the decision support tool development on the Utility function model allows the decision makers, i.e., Network Managers, IT Managers, Systems and Cloud Managers, and Senior Managers from the Finance departments, to make informed and strategic decisions in terms of the optimum solution for the transformation from a traditional campus to a smart campus.

## KEYWORDS

Smart Campus, advanced decision making analysis, utility function, decision support tool, stakeholders (external and internal)

## 1 Background

Since the 20th century, the internet has progressively served as a fundamental pillar of the modern information city and the society within. Today, most developed countries are capable of offering unlimited access to the Internet; an advantage that is becoming more common than before. As of April 2022, a global digital population of 5 billion internet users worldwide and 4.65 billion social media users have been reported (*Internet users in the world 2022, 2022*). Throughout this digital expansion, the introduction of contemporary technologies such as Artificial Intelligence (AI), the Internet of Things (IoT), and Machine Learning (ML) redefined the standards for urbanization and created the need for new urban planning strategies. These methods often involve conceptual projects which integrate information technologies to solve urban problems, now known as ‘Smart Cities.’

The term was first coined in Dutton, Blumler, and Kraemer’s book ‘Wired Cities: Shaping the Future of Communications’ published in 1987. The authors envisioned revolutionary environments where digital advancements provided ICT services to businesses and homes and called them “Wired Cities.” Likewise, many scholars adopted the same concept but differed in the term used. *Kitchin. (2013)* highlighted similar works defining “Smart Cities”; following *Dutton et al. (1987)*, Graham and Marvin in 1999 used the term “Cyber Cities,” Ishida and Isbister in 2000 used “Digital Cities,” Komninos in 2002 used “Intelligent Cities,” Hollands in 2008 used “Smart Cities,” and finally, Shepard in 2011 used “Sentient Cities.” Each of these terms defined a unique relationship between information technology and urban development. Recently, these labels were subsumed under the term “Smart Cities,” a term which has gained traction in business and government, as well as in academia (*Kitchin, 2013*).

Although the establishment of a smart city is expected to be extremely costly and large-scale, multiple studies have directed their focus toward the implementation of a “Smart Campus.” *Min-Allah and Alrashed. (2020)* highlighted a case study that was presented to the World Bank in 2014 by the University of Lille regarding the smart city and advocated the “Smart Campus” as an initial step towards the realization of the concept of a smart city. Likewise, *Omotaya et al. (2021)* concluded in their study that “smart campuses could act as living labs for future smartness of cities globally.” Similarly, *Ahmed et al. (2020)* underlined how the most significant sector in the development of any society and city is its education system, making it pertinent to incorporate recent technologies into current educational institutes in order to transform them into smart campuses.

The timeliness and promising results provided by the smart cities concept have led many researchers and scholars to pursue smart campuses within their architecture and framework. A strong correlation between the mobilization of educational resources and urbanism is recognized through these studies, suggesting the establishment of an educational facility that

encompasses sustainable digital technologies. The said facilities are known by the title ‘Smart Campus’ and their exact representations may vary according to every institution.

Literature reports from a number of countries and research institutes have directed their studies toward promoting smart cities and sustainable development. According to *Camargo et al. (2021)* Malta formulated the “Smart Island Strategy” between the years 2008 and 2010, followed by the United Kingdom, which announced “Unlocking Growth in Cities” in 2011, and Spain with its project “National Plan for Smart Cities” in 2015 (*Millard et al., 2014*). Whereas in 2017, countries such as the United States, Germany, and Hong Kong announced their smart city schemes under the titles of “Smart Cities and Communities Federal Strategic Plan: Exploring Innovation Together.” “Smart Cities Charter” and “Smart City Blueprint,” respectively (*Camargo et al., 2021*).

Similar to all the other developed countries above, the United Arab Emirates (UAE) soon gained popularity as an incubator for smart cities’ applications, seeking to present Dubai as the first smart city in the world (*Vinod Kumar, 2020*), while recently launching an initiative named ‘Smart Dubai 2021’ (*Smart Dubai 2021, 2018*).

There still is no comprehensive statement that clearly defines a smart campus, however, efforts have been made to interpret certain aspects of it. It has been highlighted that research primarily focuses on identifying the factors that can be used to define the term “Smart Campus” and identify its sophisticated infrastructure. However, no studies have been conducted regarding the end users’ perspectives on the factors they think are necessary for their respective campuses. As such, the literature does not report on any utility functions that help identify smart campus criteria for academic institutions. Hence, the absence of a decision support tool is the primary motivation behind this study.

Thus, the UAE is one of the leading countries in relation to the development of smart cities and has acted towards a smart and sustainable agenda. Nonetheless, the existence of a well-established smart campus within the UAE remains theoretical. This study, therefore, uses the American University of Sharjah as a case study to upgrade educational institutions. The study intends to determine the perception of different end users such as students, faculty members, administrative staff, and IT personnel regarding the most significant criteria of a smart campus.

Moreover, utility functions will be adopted to propose a decision support tool that aids the decision-making process in order to prioritize the different stakeholders’ perceptions of the most important criteria that can be considered for transforming a traditional campus into a smart campus.

## 2 Literature review

Literature reports on several studies that have directed their inquiries toward smart cities. These definitions have shifted their

focus from the “Internet” as being the “smart city” identifier’ in the 1990s, to a wider context that encompasses socio-economical as well as technical factors. Ramirez (2017) in her study emphasizes that despite the lack of a universal definition for what constitutes “smart cities,” cities that aspire to this designation aim to fully utilize all technological advancements to cut costs, improve efficiency, offer new services, lessen their environmental impact, and foster innovation (Aguaded-Ramírez, 2017). Similarly, Caragliu et al. (2011) identified a city to be smart when “investments in human and social capital and traditional (transport) and modern (ICT) communication infrastructure fuel sustainable economic growth and a high quality of life, with a wise management of natural resources, through participatory governance.” BSI Standards Publication. (2014) considered smart cities to include an effective integration of physical, digital, and human systems in the built environment to deliver a sustainable, prosperous, and inclusive future for its citizens. Moreover, the European Parliament. (2014) described a smart city as “a city seeking to address public issues *via* ICT-based solutions on the basis of a multi-stakeholder, municipally based partnership.” These illustrations highly reflect the integrative nature of a smart city to serve societies and improve their levels of prosperity.

Among the key forces driving technology growth in the twenty-first century is the sustainability agenda, and in order to accomplish particular objectives, governments, practitioners, society, and educators must react to these advancements. Through leaner processes and more responsive environments, smart technologies have taken over the world with their smart applications to improve quality of life. While such a significant digital transformation cannot happen overnight due to its cost implications and the complexity of the processes involved, a gradual transformation would be viable.

The primary objective of sustainable development is to promote economic and social development without compromising the environment. Batagan. (2011) conducted a study on smart cities in 2011 and highlighted that city sustainability is a multidimensional concept that incorporates economic, social, and political dimensions. Batagan. (2011) also identified three urgent goals for the concept of sustainable development: a development goal that focused on enhancing quality of life, a sustainability goal of living sustainably, and goals related to investing in technological advancement and innovation.

On the other hand, educational institutes are also expected to contribute their portion in fulfilling the United Nations (UN) Sustainable Development Goals (SDGs). The three main pillars of sustainability include facilitating economic viability, environmental protection, and social equity. Higher Educational Institutes (HEIs) are mentioned specifically under the fourth SDG, which ensures inclusive and equitable quality education while also promoting lifelong learning opportunities for all. HEIs could also contribute to the remaining SDGs

through teaching, research output as well as campus initiatives, such as implementing green smart campus solutions (Moraes et al., 2020).

Thus, it is essential to understand the crucial difference between traditional and smart campuses. Where (Yi et al., 2021) highlighted the major difference exists in the modes of learning with traditional campuses entirely dependent on a synchronous mode of learning (limited to real-time interaction among the instructors and students) in contrast with the asynchronous mode of learning offered by the smart campus that allows the students to access materials on their own pace and allow unlimited interaction. The benefit of a smart campus is not only limited to the mode of learning but offers an interconnected environment of multiple features around the campus.

The rest of this section, therefore, looks to gain a deeper understanding of smart campus concepts and definitions, identifying the criteria that underpin the formation of a smart campus while proposing a suitable approach for developing a Multi-Attribute Utility Function that will help develop a smart campus decision support tool.

## 2.1 Smart campus

A review was recently published in April 2022 which revises the technology adoption factors for IoT-based smart campuses, and it lists the efforts that have been made regarding this topic (Sneel et al., 2022). These efforts include a knowledge management model for smart campuses, which was conducted in Indonesia (Hidayat and Senses, 2022), a study on the key performance indicators for smart campuses and microgrids (Alrashed, 2020), and another study regarding an IoT-based hybrid renewable energy system for smart campuses (Eltamaly et al., 2021). The review also highlighted a methodology proposal for smart campuses (Pupiales-Chuquin et al., 2021), as well as a roadmap to smart campuses based on the IoT (Pandey et al., 2020).

Furthermore, a study on a smart campus using the Internet of Things (IoT) highlighted that the main concept of a smart campus is to develop premises that utilize resources efficiently and deliver high-quality services to the campus community while minimizing operational costs (Alghamdi and Shetty, 2016). It also elaborated on the benefits of establishing a smart campus, which included providing an interactive and creative environment for students and faculty, promoting smart energy management, bringing effective surveillance systems and real-time incidents warnings, automating maintenance and business processes, maintaining efficient parking and access control management, and providing secure payments and transparent voting systems (Alghamdi and Shetty, 2016).

A general outlook on what could be used to define a ‘smart campus’ remains necessary. Carames and Lamas (2019)

highlighted in their study that the concept of a smart campus refers to the hardware and software required to provide advanced intelligent context-aware services and applications to different stakeholders. Martins et al. (2021) described a smart campus as one where equipment and devices, apps, and people are securely connected to provide new experiences and utilities and to enhance operational effectiveness and efficiency. Petcovici and Stroulia. (2016) defined a smart campus as one where physical and virtual spaces interact through the use of intelligent devices. Likewise, Dong et al. (2020) described it as “an educational environment that is penetrated with enabling technologies for smart services to enhance educational performance while meeting stakeholders’ interests, with broad interactions with other interdisciplinary domains in the smart city context.”

These definitions indicate that the existence of a sophisticated infrastructure remains critical, and there is a need to search for and define the infrastructure on which smart campuses can operate. Moreover, despite the basic construction of campus network infrastructure being completed, Guo and Zhang. (2015) argue that traditional construction cannot meet the development of campus scientific decisions and teaching Informa tionization. However, recent technological advancements coupled with advances in data analytics tools and platforms present opportunities to transform campuses in new ways, prompting many universities worldwide to develop strategies for a smart campus (Guo and Zhang, 2015). Furthermore, the authors (2015) highlight that in order to achieve dynamic expansion,

resource adaptation, sizing deployment, and united management of a virtual data center, IoT must be incorporated into existing facilities, as well as cloud and virtualization technologies.

As such, the literature contains multiple studies that propose IoT and cloud computing as the primary foundations of smart campuses. IoT is simply a system of interrelated computing devices that are integrated with unique identifications that enable sending and receiving data over a network. The potential benefits of implementing IoT technology in smart campuses mostly revolve around three elements (Dong et al., 2020). First, it offers educators the information platform they need to monitor students’ progress and make wise decisions. Second, IoT automates smart campus operations and streamlines the teaching and learning process. Third, students’ ability in their learning activities can be tracked via IoT-based emotion or psychological recognition, and appropriate action can be taken (Dong et al., 2020). Moreover, cloud computing uses network servers on the Internet to store and manage data rather than local servers or computers. Cloud computing could be regarded as a development of distributed computing, parallel computing, and grid computing (Guo and Zhang, 2015). Through it, massive, highly virtualized computing and storage resources are combined into one large resource pool that can be utilized to deliver unified services (Guo and Zhang, 2015).

Similarly, Sutjarittham et al. (2018) proposed an overall architecture of a smart campus that accommodates use cases such as classroom attendance, student study space usage, parking

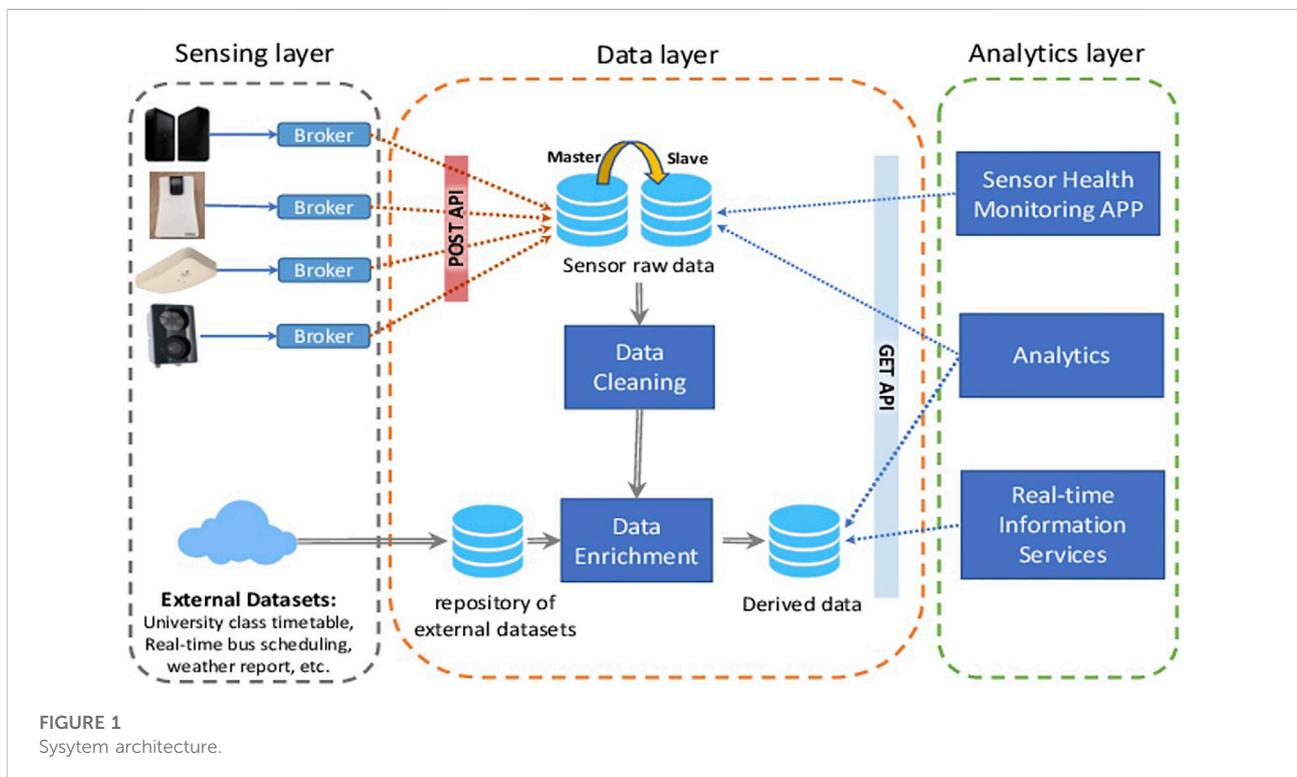
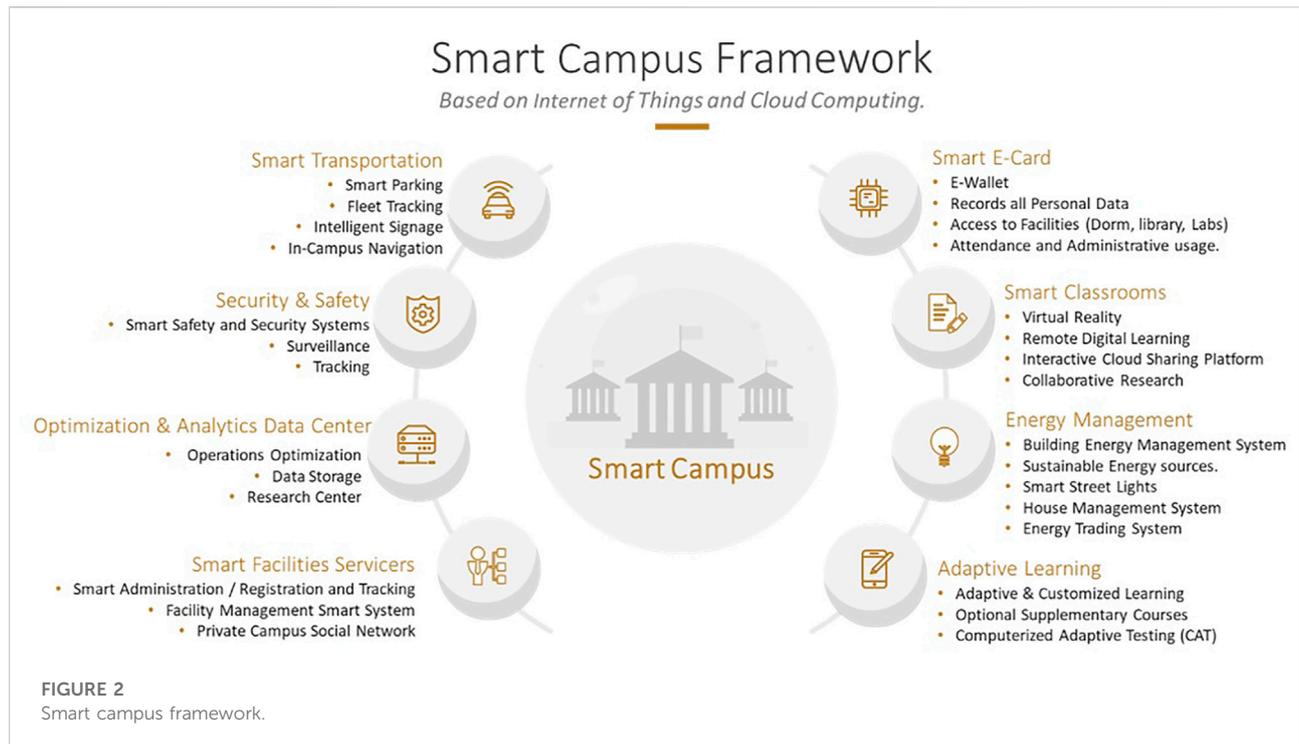


FIGURE 1 System architecture.



lot occupancy, and bus-stop wait times. The primary goal of the system architecture, according to the study's authors, is to support the collection of data produced by various smart sensing devices and its retrieval for use by applications (Sutjarittham et al., 2018) as shown in Figure 1. Figure 1 illustrates a high-level system architecture that depicts how data moves from sensors in the sensing layer to applications in the analytics layer.

This architecture was divided into three layers: A sensing layer, a data layer, and an analytics layer. The sensing layer incorporates arbitrary sensors and IoT technologies—connected devices with heterogeneous power and communications requirements into the system, the data layer stores the collected sensor values, and lastly, the analytics layer transforms sensing data into insights and actionable intelligence for applications use, enabling various services offered in a smart campus (Sutjarittham et al., 2018).

Similarly, a study by Ahmed et al. (2020) identified a set of comprehensive criteria for a smart campus, which was evaluated as to whether they formed part of a smart campus and evaluated these criteria from the stakeholders' perception, using the American University of Sharjah as a case study. The study proposed a conceptual framework for a smart campus that underpins the most important criteria from the end users' perspective. As part of their proposed framework, the criteria included applications that offered seamless and connected environments backed by the Internet of Things (IoT), cloud computing, and big data analytics (Ahmed et al., 2020).

The unified framework for smart campus consisted of 8 main and 25 application-based criteria on IoT and Cloud computing platforms (Ahmed et al., 2020). The proposed comprehensive framework is shown in Figure 2.

This paper builds on this study by using the identified criteria such as;

### 2.1.1 Smart Card

A smart card can be considered a minicomputer that uses a microprocessor chip to store and process data. The smart card makes it possible to access amenities on a smart campus quickly, easily, and securely.

### 2.1.2 Smart classroom

A smart classroom can be seen as a technology that bridges the gap between tele-education and traditional classroom settings. The idea of a smart classroom is built on contemporary technologies like 3G, 4G, the Internet of Things (IoT) platform, and other technologies to create a linked and coordinated environment (Fernández-Caramés and Fraga-Lamas, 2019).

### 2.1.3 Energy management

A system that plans and integrates power usage with nearby renewable energy sources and battery banks. It allows for storing and transferring data, which can be utilized as a data center to forecast energy output and consumption. It is a significant concept that relates to the Sustainable

Development Goals to ensure clean energy and thus should be addressed in transforming from a traditional campus to a smart campus.

#### 2.1.4 Adaptive learning

Adaptive learning is considered an educational approach based on a computerized algorithm that allows for customizing the educational strategy or learning environment in accordance with an individual's demands.

#### 2.1.5 Smart transportation

An essential aspect of facilities management is parking and fleet management. The criteria include logistics optimization, informative and quick notifications, and allows for better mobility.

#### 2.1.6 Security and safety

The criteria rely on the idea of root cause analysis (RCA), which is a methodical way to discover the "root cause" of a problem and prevent it from occurring in the future, and the security and safety standards help with advanced protection.

#### 2.1.7 Optimization and analytics data center

A vital part of any facility is the optimization and analytics data center. Consequently, the smart campus framework must address this essential criterion. The criteria support classification, data lakes, and current upgrades.

#### 2.1.8 Smart facilities and services

Dynamic campus life, responsive buildings, and easy access to athletic fields, student centers, libraries, and restaurants are all provided for stakeholders by the smart facilities services.

The study adopts the multi-criteria approach to develop a utility function that aids in the decision-making process while prioritizing the stakeholders' perceptions of the most important elements of transforming a traditional campus into a smart campus.

## 2.2 Multi-Attribute Utility Function

Multi-Attribute Utility Theory (MAUT) is the most popular Multi-Criteria Decision Making (MCDM) technique. The MAUT, which is an extension of Multi-Attribute Value Theory (MAVT), is known as the most rigorous approach to include risk preferences and uncertainty in multi-criteria decision-support strategies. The MAUT approach integrates expected utility theory to determine the best and optimum solution in a given scenario by allocating a utility to each possible attribute and measuring the best possible utility (Orsborn et al., 2009).

Over the years, several studies have adopted MAUT to support decision analysis in practical settings, for example,

Orsborn et al. (2009) adopted the utility function to estimate customer preferences over the aesthetic space. The study integrates the customer preference of the attributes in the utility function to determine the higher utility product for form generation and modification or design verification.

Moreover, Torrance et al. (1996) adopted a multi-attribute preference-based utility function to assess the health-related quality of life by integrating seven attributes such as sensation, mobility, emotion, cognition, self-care, pain, and fertility. The Health Utilities Index adopted a standard gamble technique to measure the utilities.

Likewise, the study by Abolbashari et al. (2018) developed a model based on the utility function for procurement performance management by adopting a case study. The decision model allows the decision makers to determine the optimum Key Performance Indicators (KPIs) to invest in that have an impact on procurement performance. The findings of the study indicate that the optimum solution for the organization is to invest and allocate its resources to staff training and procurement cycle time to achieve the highest utility.

Another application of the utility function was identified by Walsh et al. (2004). The study adopted the utility function for autonomic resource allocation among multiple applications. For the estimation of the utilities, the study adopted and compared two methodologies based on the queuing-theoretic performance model and model-free reinforcement learning. Furthermore, to overcome the issue of outsourcing contracts for vendor selection, the study by Teixeira de Almeida. (2007) adopted the utility function and ELECTREE function to propose a model that integrates multicriteria evaluation. The study illustrates the effectiveness of the model through a case study. Furthermore, the utility function was presented in the study by Jæger. (2007), which intended to determine the utility of both economic and social returns to education. The findings of the study concluded that economic returns to education are somewhat more important than social returns.

Thus, this indicates the effectiveness of adopting the utility function to determine an individual's preferences based on multiple attributes and criteria. The utility function allows us to determine the rational choice theory to analyze human behavior and preferences. Besides, the major benefit of adopting MAUT is that it considers the uncertainty, which is not accounted for in many MCDM methods.

From the literature, it can therefore be summarized that the definition of the smart campus can be underpinned by several criteria while the importance of each criterion and/or sub-criterion vary according to every institution. Thus, these can be used to address the research question of which factors are perceived to be important by the end user and demonstrate how MAUT can be adopted to develop a decision support tool that can aid the decision maker to make informed decisions.

This study, therefore, intends to adopt the AHP and utility function by integrating the multiple criteria of a smart campus to

determine the optimum criteria/solution to invest in to transform a traditional campus into a smart campus.

### 3 Methodological steps

This section presents the following methodological steps adopted to achieve the aim of the study which is to develop a decision support tool that allows the decision-makers to invest in the optimum solution to transform a traditional campus into a smart campus.

#### 3.1 Stage I literature review

This stage helped understand “Smart Cities and Smart Campuses” and identifies the underpinning criteria that define a smart campus in order to adopt the best solution to develop decision support tool that can be adopted by the decision-makers to help identify the most important criteria for the transformation of a traditional campus into a smart campus.

#### 3.2 Stage II surveys

A survey is conducted targeting a sample of students, faculty, administrative staff, and IT support personnel from across the university to evaluate their perceptions of the most significant criteria of a smart campus in AUS, using AHP analysis based on pairwise comparisons to compare the perception of students, faculty, and administrative staff.

#### 3.3 Stage III decision support tool

This stage helps with developing the decision support tool based on the utility function. The stage allows the decision-makers (university administrators) to rank the smart campus criteria based on their level of impact and their importance in transforming a traditional campus into a smart campus. In addition, the decision-makers will be asked to rank each criterion on a scale from -10 to 10, this allows us to determine the acquired utility and the optimum decision or criteria to invest.

## 4 Data collection and analysis

This section presents the data collection methods and analysis adopted to develop a decision support tool that allows the decision-makers to determine the optimum decision or criteria to invest in transforming a traditional campus into a smart campus.

### 4.1 Case study

The study targets the American University of Sharjah as a case study due to its forward-looking approach that endeavors to continuously move with the times and keep up with current developments as well as respond to market needs through the provision of excellent student services and environment. The university is a leading institute in the region with the reputation of being a research hub. The current research projects by the institute such as developing a virtual charging system for electric vehicles, and the adoption of IoT and big data to support UAE-wide smart city initiatives, shows the commitment and role of the institute in the development of smart cities, in general, and the smart campus, particularly.

### 4.2 Pre-development phase

The phase includes a survey, which was developed by adopting an analytical hierarchy process (AHP) to conduct a pairwise comparison among the 8 smart campus criteria and their corresponding 25 applications, as shown in Figure 2. The study adopted purposive sampling by targeting students, faculty, administrative staff, and IT support personnel, which the authors identified as being actively involved in research of the latest technologies and having sound knowledge of smart cities. The survey intended to identify the most important criteria based on the perceptions of a diverse group of end users to transform a traditional campus into a smart campus. The survey also drew upon a comparison among the different participant groups to determine which criteria and applications are important to each group. The intent of the comparison was to set a benchmark for the decision-makers to make decisions based on the perception of end users.

TABLE 1 Participant's profile.

Demographic profile	Categories	Frequency
Position	Students (research assistants)	10 (26.3%)
	Academic faculty	08 (21.1%)
	Staff members	10 (26.3%)
	IT support personnel	10 (26.3%)
Gender	Male	28 (73.7%)
	Female	10 (26.3%)
Courses taught	PhD	02 (5.3%)
	MSc	08 (21.1%)
	Undergraduates	14 (36.8%)

TABLE 2 Saaty's Scale for AHP analysis.

Numerical value	Description
1	EQUAL IMPORTANCE
3	Slight importance of one over another
5	Moderate importance of one over another
7	Very strong importance
9	Extreme importance of one over another
2,4,6,8	Intermediate values between two adjacent values

### 4.3 Participant's profile

The survey was distributed to selected participants that have sound knowledge of the smart city and smart campus. A total of 38 participants were targeted as summarized in Table 1.

The survey participants included 10 students working in research assistant positions, 8 faculty members that taught Ph.D. MSc, and undergraduate courses, 10 staff members,

and 10 IT support personnel that specializes in the latest technologies and had a sound knowledge of smart cities. To determine the perception of the participants belonging to different groups, an online survey was sent to the selected participants. The survey asked the participants to pairwise rank the 8 criteria and 25 applications using a Saaty scale as shown in Table 2.

The findings of the survey were presented in the table below:

### 4.4 Students

The survey response accumulated from the students' participants were shown in Table 3. The findings suggested that smart security and safety, followed by adaptive learning, were the preferred criteria that the students perceived to be suitable for the transformation of a traditional campus into a smart campus.

Moreover, the most important application perceived by the students to transform the campus into a smart campus was collaborative research, e-wallet payment, and in-campus navigation.

TABLE 3 Student's perception regarding significant smart campus criteria.

Criteria	Weight	Rank	Applications	Weight	Rank
Smart card	0.0599	8	Attendance	0.1615	18
			For dorms	0.0828	25
			For library usage	0.1305	24
			E-wallet (payments)	0.3963	2
			To record all personal data	0.2286	13
Smart classroom	0.0954	7	Virtual reality	0.1435	21
			Remote digital learning	0.1403	22
			Interactive cloud sharing	0.2940	8
			Collaborative research	0.4219	1
Energy management	0.1085	5	Buildings energy management system	0.1499	20
			Sustainable energy	0.3269	6
			Smart street lights	0.1529	19
			House management system	0.2059	14
			Energy trading system	0.1641	17
Adaptive learning	0.1479	2	Adaptive learning	0.3893	4
			Optional supplementary courses	0.3370	5
			Computerized adaptive testing (cat)	0.2736	9
Smart facilities services	0.1048	6	Smart safety & security systems	0.2610	11
Optimization and analytics data centers	0.1215	4	Optimization & analytics data center	0.1985	16
Security and safety	0.2160	1	Smart facilities services	0.2943	7
			Private campus social network	0.2460	12
Smart transportation	0.1455	3	Smart parking	0.2042	15
			Fleet tracking of all campus transportation	0.1315	23
			Intelligent signage	0.2734	10
			In-campus navigation	0.3908	3

## 4.5 Staff members

Furthermore, the survey was sent to selected members of staff who are involved in dealing with smart technologies. The survey findings suggested that both, the students, and the staff members perceived smart security and safety followed by adaptive learning as the most important criteria for transformation into a smart campus, as shown in [Table 4](#).

Furthermore, the applications perceived to be important to the staff members were adaptive learning, collaborative research, and optional supplementary courses.

## 4.6 Academic faculty

Moreover, the survey also targeted academic faculty that taught Ph.D. MSc, and undergraduate courses related to smart cities and big data. The findings of the survey in [Table 5](#), show that the faculty perceive that smart transportation, followed by smart security and safety, were the most significant criteria to transform a traditional campus into a smart campus.

Whereas, for the applications, the faculty perceived optional supplementary courses, computerized adaptive testing (CAT), and in-campus navigation to be the most significant to invest in. [Table 6](#).

## 4.7 IT support personnel

Finally, the survey targeted IT Support personnel that play a key role in transforming the institute into a smart campus. The findings suggested that the IT personnel perceived the adoption of smart classrooms and smart security and safety as the most important criteria to transform a traditional campus into a smart campus.

The applications deemed important by the IT department were adaptive learning, computerized adaptive testing (CAT) and in-campus Navigation.

To conclude, there existed a consensus among the perception of a diverse group of participants who perceived *smart security and safety*, *in-campus navigation*, and *adaptive learning* as the most important criteria and applications to transform a traditional campus into a smart campus. However, there also existed a difference in the perception of the importance of criteria such as *smart transportation*, *smart classroom*, and applications

TABLE 4 Staff members' Perception regarding Significant Smart Campus Criteria.

Criteria	Weight	Rank	Applications	Weight	Rank
Smart card	0.07917	8	Attendance	0.2201	16
			For dorms	0.1440	24
			For library usage	0.2039	20
			E-wallet (payments)	0.1681	23
			To record all personal data	0.2635	8
Smart classroom	0.1176	5	Virtual reality	0.1898	21
			Remote digital learning	0.2400	13
			Interactive cloud sharing	0.2563	9
			Collaborative research	0.3137	2
Energy management	0.0835	7	Buildings energy management system	0.2235	15
			Sustainable energy	0.2049	18
			Smart street lights	0.1388	25
			House management system	0.2448	12
			Energy trading system	0.1878	22
Adaptive learning	0.1590	2	Adaptive learning	0.4112	1
			Optional supplementary courses	0.3127	3
			Computerized adaptive testing (cat)	0.2760	6
Smart facilities services	0.1194	4	Smart safety and security systems	0.3068	4
Optimization & analytics data centers	0.1276	3	Optimization and analytics data center	0.2087	17
Security and safety	0.2093	1	Smart facilities services	0.2800	5
Smart transportation	0.1041	6	Private campus social network	0.2043	19
			Smart parking	0.2666	7
			Fleet tracking of all campus transportation	0.2334	14
			Intelligent signage	0.2515	10
			In-campus navigation	0.2484	11

TABLE 5 Academic faculty perception regarding significant smart campus criteria.

Criteria	Weight	Rank	Applications	Weight	Rank
Smart card	0.0915	8	Attendance	0.1673	22
			For dorms	0.1359	25
			For library usage	0.1955	18
			E-wallet (payments)	0.2313	13
			To record all personal data	0.2698	8
Smart classroom	0.1191	5	Virtual reality	0.1881	19
			Remote digital learning	0.2683	9
			Interactive cloud sharing	0.2483	11
			Collaborative research	0.2951	5
Energy management	0.1126	7	Buildings energy management system	0.1576	24
			Sustainable energy	0.2456	12
			Smart street lights	0.1609	23
			House management system	0.2681	10
			Energy trading system	0.1675	21
Adaptive learning	0.1193	4	Adaptive learning	0.2800	7
			Optional supplementary course	0.3906	1
			Computerized adaptive testing (cat)	0.3293	3
Smart facilities services	0.1207	3	Smart safety & security systems	0.2876	6
Optimization & analytics data centres	0.1171	6	Optimization & analytics data centre	0.2048	17
Security and safety	0.1450	2	Smart facilities services	0.2980	4
			Private campus social network	0.2094	15
Smart transportation	0.1743	1	Smart parking	0.1797	20
			Fleet tracking of all campus transportation	0.2066	16
			Intelligent signage	0.2299	14
			In-campus navigation	0.3835	2

such as *computerized adaptive testing (CAT) and optional supplementary courses*. But it can be suggested that the criteria and applications relate to the scope of the group due to which they perceived it to be important.

The findings from the survey can be adopted as a benchmark and reference for the decision-makers to make decisions that transform a traditional campus into a smart campus.

## 4.8 Development phase

This phase targets Network Managers, IT Managers, Systems and Cloud Managers, and Senior Managers from the Finance departments as the decision-makers. The phase includes the following stages:

Stage I—Firstly, the decision-makers were presented with a survey to determine the level of impact of the criteria to transform a traditional campus into a smart campus. For this purpose, the participants were asked to rank the level of impact of each criterion based on a three Likert scale of level of impact such as *Low, Medium, and High* based on their expertise and

perception. To determine the probability distribution, an equation of probability was adopted (Abolbashari et al., 2018):

$$P(I_i^j), \text{ where } \sum_1^j P(I_i^j) = 1$$

For this scenario,  $j = 3$  (Low, Medium, and High);  $i = 8$  smart campus criteria,  $k = 1$  (the study's intent to determine the optimum solution to invest).

The total probability distribution of the response of the experts was then aggregated in a form of probability. This allows for the integration of the fuzzy nature of the criteria and its impact perception in the decision. The findings of the level of perceived impact of the criteria on transforming a traditional campus into a smart campus are shown in Table 7.

The results in Table 7 can be interpreted as 39% of the experts perceived a smart card has a low impact or plays an insignificant role in transforming a traditional campus into a smart campus, while 31% and 30% of the participants believed the criterion has a medium to high impact and role in transforming traditional campus to a smart campus. Furthermore, it can be seen that 58% of the participants perceived the impact level of smart security

TABLE 6 IT support Personnel’s perception regarding significant smart campus criteria.

criteria	Weight	Rank	Applications	Weight	Rank
Smart card	0.0934	8	Attendance	0.1546	25
			For dorms	0.1830	22
			For library usage	0.2005	20
			E-wallet (payments)	0.2282	15
			To record all personal data	0.2336	12
Smart classroom	0.1254	2	Virtual reality	0.2081	17
			Remote digital learning	0.2676	7
			Interactive cloud sharing	0.2662	8
			Collaborative research	0.2578	9
Energy management	0.1114	5	Buildings energy management system	0.1884	21
			Sustainable energy	0.2757	6
			Smart street lights	0.1759	23
			House management system	0.2021	19
			Energy trading system	0.1577	24
Adaptive learning	0.1153	4	Adaptive learning	0.3860	1
			Optional supplementary courses	0.2880	4
			Computerized adaptive testing (cat)	0.3259	3
Smart facilities services	0.1091	6	Smart safety & security systems	0.2780	5
Optimization and analytics data centres	0.1226	3	Optimization & analytics data centre	0.2322	14
Security and safety	0.2183	1	Smart facilities services	0.2440	11
			Private campus social network	0.2455	10
Smart transportation	0.1040	7	Smart parking	0.2050	18
			Fleet tracking of all campus transportation	0.2329	13
			Intelligent signage	0.2225	16
			In-campus navigation	0.3394	2

TABLE 7 Total probability distribution.

Criteria\ impact level	Low	Medium	High
Smart card	0.39	0.31	0.30
Smart classroom	0.19	0.38	0.43
Energy management	0.23	0.43	0.35
Adaptive learning	0.20	0.32	0.48
Smart facilities services	0.25	0.43	0.32
Optimization and analytics data centres	0.28	0.39	0.33
Security and Safety	0.05	0.37	0.58
Smart transportation	0.23	0.37	0.40

and safety as high to transform a traditional campus into a smart campus. The findings align with the findings of AHP analysis as conducted in the predevelopment stage.

Stage II—Secondly, to determine the optimum solution and criteria that the institution should invest in to transform a traditional campus into a smart campus, the decision-makers were asked to assign a utility to each criterion on a scale from -10 to 10. Each criterion utility is assigned a conditional

probability distribution that depends on the influence node  $I_i^j$  and decision node  $D_i^k$ . For each criterion, the utility node is represented by  $U(I_i^j, D_i^k)$  as shown in Table 8 below:

The table shows the utility of each criterion against the impact factor of the criteria to transform a traditional campus into a smart campus.

Stage III—Finally, the expected utility for the decision for each criterion can be determined by adopting the utility equation as adopted by Abolbashari et al. (2018).

$$EU(D_i^k) = \sum_j P(I_i^j) U(I_i^j, D_i^k)$$

Here D, Decision node; U, Utility node; I, Influence node; i, number of criteria; j, number of states in criteria, k: to invest,  $D_i^k$ : decision to invest,  $I_i^j$ : impact j of criteria on transformation to smart campus and  $U_i^k$ : the utility associated with  $I_i^j$  and  $D_i^k$

The equation calculates the expected utility of each criterion and allows us to determine the optimum solution and criteria to invest in to transform a traditional campus into a smart campus. The findings of the expected utility and optimum solution are presented in Table 9 below.

TABLE 8 Utility associated with decision to invest and state of criteria.

	Smart card	Smart classroom	Energy management	Adaptive learning	Smart facilities services	Optimization & analytics data centers	Security & safety	Smart transportation
Low	-7	-7	-7	-6	-6	-6	-7	-7
Medium	4	3	4	4	5	3	3	4
High	9	9	9	9	8	8	9	7

TABLE 9 Optimum decision.

Criteria		Expected utility	Rank	Optimum decision
Smart card	$D_1^1$	1.3807245	8	$D_7^1$
Smart classroom	$D_2^1$	3.4102705	3	
Energy management	$D_3^1$	3.1137731	4	
Adaptive learning	$D_4^1$	3.9608553	2	
Smart facilities services	$D_5^1$	3.0935223	5	
Optimization and analytics data centers	$D_6^1$	2.0645281	7	
Security and safety	$D_7^1$	5.8427126	1	
Smart transportation	$D_8^1$	3.0024836	6	

TABLE 10 Abbreviation table.

Terms	Abbreviations
Analytical hierarchy process	AHP
United arab emirates	UAE
Artificial intelligence	AI
Internet of things	IoT
machine learning	ML
Information and communications technology	ICT
Higher educational institutes	HEIs
United nations	UN
Sustainable development goals	SDGs
Multi-attribute utility theory	MAUT
Multi-criteria decision making	MCDM
Computerized adaptive testing	CAT
Elimination and choice expressing reality	ELECTREE
Technique for order of preference by similarity to ideal solution	TOPSIS

According to the results, from Table 9, the optimum decision to invest is smart security and safety followed by adaptive learning. The findings of the decision support tool also align with the perception of end users determine through AHP analysis in the predevelopment phase.

## 5 Discussions

The smart campus is an emerging trend that allows institutions to integrate smart technologies with their physical infrastructure to provide better services, decision-making, and

campus sustainability, among other things. In recent years, an exponential interest in the features and implementation of these features with the latest technologies have been discussed in the literature to define a “Smart Campus” as mentioned in [Table 10](#). However, with technological development on the rise, there is no single best definition or platform that denotes a smart campus. Thus, this is an emerging phenomenon that depends on multiple criteria and features.

A number of studies have proposed various platforms with multiple features and applications to transform a traditional campus into a smart campus such as cloud computing, the Internet of Things (IoT), Augmented reality (AR), and artificial intelligence to name a few as background platforms. These studies ([Muhamad et al., 2017](#); [Dong, et al., 2020](#); [Min-Allah and Alrashed, 2020](#)) also propose multiple human-centered and learning-centered features and applications, and environmental services including smart grid, learning environment, waste and water management, intelligent building, health and fitness, transportation, and smart parking.

These transformational applications were adopted by KFUPM by integrating ultrasonic sensing technologies and a database system that allows access control to dorms and assigns parking spots in the system to those living in dorms based on their preferences. In addition, this parking guidance system helps the students to park effectively ([Alghamdi and Shetty, 2016](#)).

Likewise, the Massachusetts Institute of Technology (MIT, Cambridge), in partnership with Microsoft, developed collaborative research with the goal of implementing an intelligent campus known as the MIT iCampus. The iCampus allows the end users to benefit from a class communicator system (CSS) and a class learning partner (CLP) to overcome the issues of miscommunication between instructors and students. The system offers an enhanced learning experience by offering instant feedback assessment and analysis ([Alghamdi and Shetty, 2016](#)).

Similarly, New York University (NYU) adopted the smart card to allow the stakeholders such as students, faculty, and staff to have a seamless experience around the campus. The card allows the end user to access the buildings, labs, and dorms. Moreover, the card acts as a debit card that can be used for payments for books, food, and other facilities around the campus. Thus, this shows the adoption of technologies allows for seamless and efficient management and experience ([Alrashed, 2020](#)).

However, where a number of these studies ([Abuarqoub et al., 2017](#); [Sánchez-Torres et al., 2018](#); [Imbar et al., 2020](#)) propose the latest technologies and features to build a smart campus, they fail to include the end user’s perception of what is important to their experience on a smart campus. This research, therefore, targets a diverse range of stakeholders by conducting a survey with a sample of students, faculty, administrative staff, and IT support personnel from the leading institute in the region as a single case study.

At the micro level, the findings suggested that smart security and safety, followed by adaptive learning, were the preferred criteria that the students perceived to be suitable for the transformation of a traditional campus to a smart campus. Moreover, the most important application perceived by the students to transform the campus into a smart campus was collaborative research, e-wallet payment, and in-campus navigation. Likewise, for the staff members, a similar perception was witnessed. The survey findings suggested that both the students and the staff members perceived smart security and safety, followed by adaptive learning, to be the most important criteria for transformation to a smart campus. Whereas the applications perceived to be important to the staff members were adaptive learning, collaborative research, and optional supplementary courses. Moreover, for the faculty, the findings suggested that the faculty perceived that smart transportation, followed by smart security and safety, were the most significant criteria to transform a traditional campus into a smart campus. Whereas, for the applications, the faculty perceived optional supplementary courses, computerized adaptive testing (CAT), and in-campus navigation to be the most significant to invest in. Finally, at the micro level for the IT personnel, the findings suggested that the IT personnel perceived the adoption of smart classrooms and smart security and safety as the most important criteria to transform a traditional campus into a smart campus.

Thus, at a macro level, it can be concluded that a consensus in the perception of a diverse group of participants was witnessed as they all perceived smart security and safety, in-campus navigation, and adaptive learning as the most important criteria and applications to transform a traditional campus into a smart campus. However, there also existed a difference in the perception of the importance of criteria such as smart transportation, smart classroom, and applications such as computerized adaptive testing (CAT) and optional supplementary courses. But it can be suggested that the perceived importance of criteria and applications is related to the scope of the group.

Secondly, the development phase for the decision support tool based on the Probability distribution and Utility Function, targeting the experts i.e., Network Managers, IT Managers, Systems and Cloud Managers, and Senior Managers from the Finance departments as the decision-makers to make informed and strategic decision in terms of the optimum solution for the transformation of a traditional campus into a smart campus, suggested that 30 percent of the experts thought the smart card would have a high impact and play a significant role in transforming traditional campus to a smart campus, where 43 percent thought smart classroom, 48 percent thought adaptive learning and the highest percentage of 58 percent of the experts and decision-makers thought smart security and safety would have the

highest impact when it comes to transforming a traditional campus to a smart campus.

## 6 Conclusion and future work

The increasing importance of smart technologies has led cities to move faster towards implementing smart solutions to improve the living standards of individuals and the quality of their lives. Good education forms one of the most important factors to develop societies as it plays a vital role in building future generations and cultivating the way they think. Thus, this results in many universities and research institutes directing their studies toward smart cities and sustainable development. With United Arab Emirates (UAE), in general, and Dubai, in particular, being an incubator for smart cities' applications with their initiative named "Smart Dubai 2021," the development of smart campuses has gained popularity.

This study, therefore, lays the groundwork for institutions and decision-makers to make decisions by adopting a decision support tool based on the utility function that encompasses the most important criteria for promoting a smart campus, using AUS as a case study. The study identifies and compares the perception of diverse end-user groups such as students, administrative staff, academic faculty, and IT support personnel in terms of the most significant criteria for a smart campus. Finally, the tool allows the decision makers (university administrative staff) to make decisions by adopting DST to determine the optimum solution/criteria to invest in to transform a traditional campus into a smart campus. The limitation of the study, however, is the survey from a single institution in the United Arab Emirates region. Therefore, the findings are limited to end-user perception in the region. However, the tool proposed can be adopted by other researchers to implement a similar methodology. In addition, the scope of the study can be increased by taking into account multiple institutions within and outside the GCC and drawing on the comparison of the end user's perception by adopting the TOPSIS method (Technique for Order of Preference by Similarity to Ideal Solution).

The findings of the study will allow the presentation of suitable sustainable practices that can enhance the sustainability rating of university campuses in the sociological, environmental, and economical aspects. This can further expand to positively affect the sustainability awareness inside cities and improve the living standards for individuals. The dissemination of the findings will open opportunities for further research and will attract researchers to study the multidisciplinary aspects of a smart campus to make a real impact through innovative research. Lastly, the findings will have a positive impact on the quality of education and socio-environmental preservations, as well as

create pathways for future smart applications to take place on many other campuses and in future cities.

## Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

## Ethics statement

The studies involving human participants were reviewed and approved by Institutional Review Board, American University of Sharjah. The patients/participants provided their written informed consent to participate in this study.

## Author contributions

All authors contributed to the study's conception and design. Material preparation, data collection, and analysis were performed by VA, SS, NA, and MG. The first draft of the manuscript was written by SS and all authors commented on previous versions of the manuscript. All authors read and approved the final manuscript.

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## Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest

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