

## Review Article

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# Using IoT and AI to replenish household food supplies: A systematic review

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**Abstract.** Food wastage because of the lack or incompleteness of a household replenishment system is an essential topic to be addressed. An appropriate utilization of Internet of Things (IoT) and Artificial Intelligence (AI) technologies with particular components is needed to design a smart household replenishment system to reduce food wastage. Therefore, this systematic review is dedicated to survey papers utilizing IoT and AI tools for perishable items storage compartments, as they are always full of items that need to be monitored. This study was conducted by following the PRISMA search strategy. It examined 70 papers in chronological order starting from 2000 when LG Electronics invented the first smart refrigerator, and research on technology involvement in food storage compartments increased. This comprehensive research aims to point out the approaches, contributions, used components and limitations of the reviewed papers to develop a unified framework for a household replenishment system. The analysis resulted in 43 approaches using IoT technology, 27 using AI, and recently the use of AIoT has been trending in the past two years. This systematic review provides future directions for researchers acquired from the limitations of the reviewed papers to enhance the household replenishment system by developing and adding required features in smart food storage compartments. Further investigation into smart home appliances would lead to extensive approaches like smart shops, industries, and eventually smart cities.

**Keywords:** Household Replenishment System, intelligent refrigerator, artificial intelligence (AI), internet of things (IoT), food wastage

## 1. Introduction

The waste of food is an essential issue to consider. There are many reasons for the waste. Without an organized shopping list, consumers purchase food based on imperfect memories of what remains at home and advanced marketing techniques used by supermarkets and other vendors. Additionally, severe weather conditions (or a pandemic in some cases) that require home isolation may make people unintentionally buy more than needed. People may forget their uneaten food, be unaware of expiration dates, store leftover food for too long, and food not well orga-

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nized inside the refrigerator—losing food has many consequences. As part of home appliances, this paper intended to focus on refrigerators in particular as they always have perishable items that require monitoring. In contrast, other home appliances, such as washers and dryers, are occasionally used.

Inside the refrigerator, wasted food will unnecessarily consume space and electricity; the spoiled food means the water, labor, and other resources needed to produce it have been wasted [6,20,34,41,52,61]. A specialized agency of the United Nations called the Food and Agriculture Organization (FAO) studied food that ends up being wasted worldwide. The FAO study shows that one-third of the total amount of food worldwide is thrown away or lost, and consumers cause 40% of the waste [20,94]. The study was based on related kitchen appliances and the storage used to store food [20]. Adding features that track stored food items, monitor consumption and remaining food levels, will reveal a pattern of household behavior and consumption. The pattern will form the household replenishment system that will provide a shopping list to guide the household and stores. The historical information regarding the consumption rate of each perishable item will be gathered and recorded by the implanted sensors in the storage compartment. Then, it will be analyzed accordingly to form the pattern and predict the upcoming purchase along with the expected quantity [53,87].

Artificial Intelligence technology's original seeds were planted in 1950 by Alan Turing [85]. Mr. Turing, proposed his idea in his book "Computing Machinery and Intelligence," wherein he suggested people be concerned about the question "Can machines think?" [85]. From the proposed question by Mr. Turing, the basic idea of AI can be understood. It allows computers and machines to process data to be useful information using AI tools and techniques such as text mining, data mining, machine learning, image recognition, image processing, voice recognition, natural language processing, and other tools that allow machines to think, make and help decision-making processes like humans [34]. For example, Shweta in 2017 presented a solution for the smart refrigerator by using Artificial Intelligence technology [81]. The presented solution uses a machine-learning approach called the Aging algorithm. It includes a database to record the data processed by a microcontroller using the algorithm. The database is pre-loaded with images of different types of vegetables to train the data. The data was trained to detect the texture, shape, and other features of the captured images. It allows the system to compare the captured images with the loaded images. The comparison takes place to make a proper decision on the vegetable's age. The system aims to monitor only vegetables. Thus, a camera is placed inside the refrigerator where the vegetables are located. The camera captures images and is connected to the microcontroller, where the images get analyzed for their age. After the analysis, the images will be sent to a microprocessor that converts and transmits the received information into signals. The signals will be received in the form of voice. Then, an attached voice indicator notifies the household about the vegetable status [81].

Nagarajua et al. [59] proposed a conceptual idea of using artificial intelligence in smart refrigerators to eliminate the wastage of food. This paper does not include specific components to be used; it was kept for the implementer to decide. The proposed solution has two approaches. One approach is using image recognition, referred to as predictive vision. Image recognition is an approach that could be applied by one of the AI tools called deep-neural networking. The authors suggested using the conventional neural network for vast data in this paper. The captured images will be processed, and the system will make the decision based on a comparison with pre-loaded identified images in a database and trained data. The second approach is using Natural Language Processing (NLP), which allows the household to communicate with the smart refrigerator by speaking out loud whenever they place items or take out items into or from the refrigerator; quantities of items can also be mentioned in the refrigerator. NLP will recognize the spoken information by the household and record it in a database. The database will be updated every time the household gives any voice command [59].

Gull et al. [27] proposed a conceptual idea for a system to enable the household to monitor food items stored in a smart refrigerator using Artificial Intelligence of Things (AIoT) via a personal computer. A novel idea of embedding the AI. Using decision tree advanced ID3 model, a machine learning algorithm, and an eNose system with different gas sensors. The gas sensors are used to identify food items, meat, rice, fruits, and vegetables from the emitted gases. Then, the captured data is sent to a trained database to label the food items accordingly [27]. Therefore, the camera module, eNose system, voice recognition, and Natural Language Processing technologies could mimic human action. The obtained data could be recorded and processed to train prediction models. Upon verification and validation of the developed models. The models would help to make proper decisions in predicting the suggested shopping lists and provide the desired output.

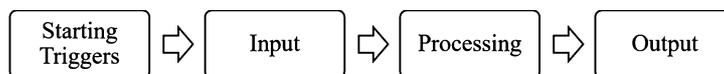


Fig. 1. Data flow of household replenishment system.

The state-of-the-art Internet of Things (IoT) technology made it easier to connect things (objects and electronic devices) to the Internet. IoT technology allows objects to communicate, exchange, and store information via communication channels, databases, and internet protocols that make up an IoT platform. The IoT is driven by AI technology. The IoT started in 1982 at the University of Carnegie Mellon. Computer science students modified a Coke machine so it would inform a computer network about the inventory status and drink temperature. The IoT revolution began in earnest in the early 1990s [21,40,84]. Just as the Carnegie-Mellon students found Coke machine data useful, recent technological developments, as shown in this literature review, will bring accurate, timely information to consumers to automate their household replenishment systems. Information can be obtained by involving the new technology in home appliances. A patent titled “Household consumable item automatic replenishment system including intelligent refrigerator” granted to Sone in 2001 demonstrated an automated household replenishment system. It included various components that are interconnected and connected to the Internet to monitor stored items’ status. A detailed diagrams and illustrations were provided to illustrate the workflow of the proposed system [83]. Thus, it is essential for an easier lifestyle to maintain household demands along with proper management of supplies in a smart structure. The potential of IoT would lead to smart homes, markets, industries, and eventually smart cities [3].

Realizing that information flow can arise after things have been identified and devices are connected in the same network, called Information integration, leads to thinking about making that possible. The literature found that exchanging data can be done in a system and stored in a shared database such as Google Firebase or an offline database. The data can be retrieved as input from sensors, RFID systems, camera modules, and barcode systems in many ways. A controller such as the Arduino UNO dashboard device can be configured to retrieve data from different sources, process them, and send them to a mobile application or a central database as an output [9], as shown in Fig. 1. The database is accessible by authorized people via the Internet. The information stored in this database can be viewed, managed, modified, and updated for several reasons. From the user’s perspective, the information could be about the flow of the stored items at home, allowing the households to make proper decisions about what to purchase next, know what items are about to expire, and control the surrounding climate to ensure freshness. Furthermore, if stores and suppliers are allowed to interact with that database, they can analyze the habits of each household, improve products, set proper marketing plans, form a pattern of subsequent purchases, and ensure the product availability [88].

Most current devices can be identified and connected to the internet wirelessly via WiFi or Bluetooth connections or wired via Ethernet for exchanging information and other purposes. Appliances such as regular refrigerators or storage cabinets require specific AI or IoT components to allow them to be converted into smart storage compartments of perishable items.

This paper will focus on AI and IoT components and information integration methods used by the reviewed papers that enhance the household replenishment system. Therefore, this paper analyzes relevant research about how AI and IoT technologies can create a unified framework that includes all used components to eliminate food waste and enhance the Household Replenishment System and provide future directions. The presented possible combination of all used components could be used as guidance for developers and future directions for researchers.

The remainder of this systematic review is organized as follows; section two is the research methodology, section three is the Household Replenishment Systems Analysis, section four is the findings and discussions, and the fifth section is the conclusion and future directions.

## 2. Research methodology

The comprehensive research has followed the PRISMA guidelines, as shown in Fig. 2, which are helpful for researchers creating and arranging systematic reviews [61].

Several factors cause the slow movement of research and implementation of the new technologies needed to convert storage compartments into smart, home storage compartments for perishable items. Among them are the cost of system configuration, manufacturing, and research and security issues [61].

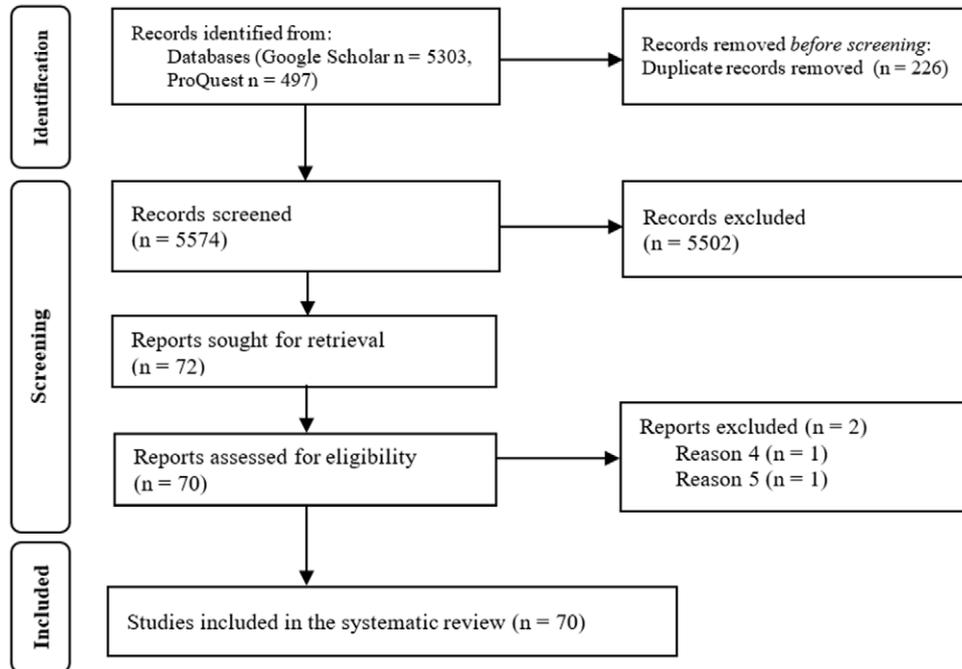


Fig. 2. Search strategy following PRISMA flow diagram 2020.

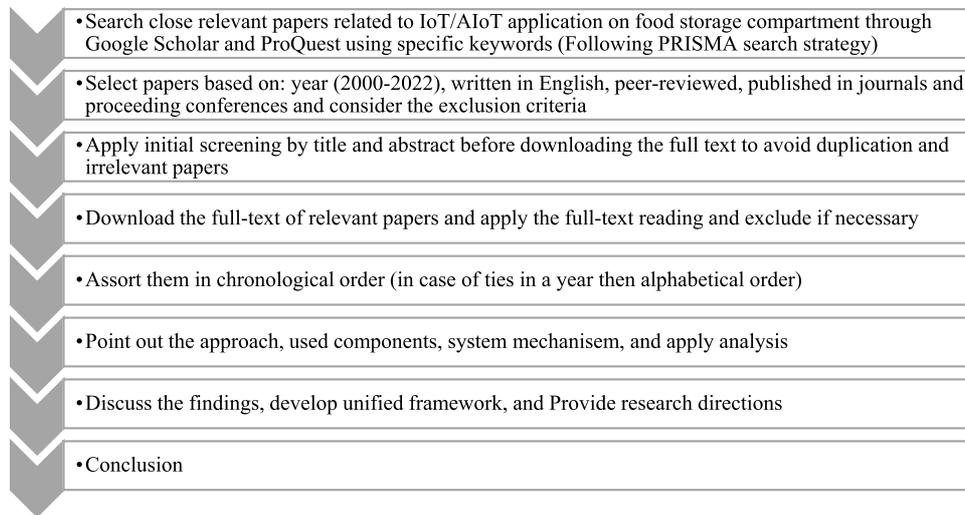


Fig. 3. Summary of the systematic review methodology.

This systematic review surveys the relevant work on an application of IoT, Artificial Intelligence of Things (AIoT) technologies on perishable items storage compartments, the household interaction, and a detailed description of components, 2) point out the advantages and limitations of the concept, 3) find the frequency of the components appear in the research, 4) and perform data mining technique that built a decision tree for a continuous variable decision on the reviewed papers, and 5) offers a unified framework to developing a system that can include all necessary components. Also, future research directions to enhance the household replenishment system to reduce food wastage will be provided, as shown in Fig. 3.

Table 1  
Summary of initial search results on both databases

Used keywords	Number of results		Total
	ProQuest	Google scholar	
“SMART refrigerator”	283	1860	2143
“Smart refrigerator AND “internet of things”	0	131	131
“Smart refrigerator” AND “artificial intelligence”	86	647	733
“SMART Fridge”	96	1790	1886
“Grocery Ordering Systems”	2	18	20
“Fridge” AND “IoT” AND “AI” AND “SMART refrigerator”	5	94	99
“smart food storage”	0	26	26
“INTELLIGENT REFRIGERATOR”	13	447	460
“INTELLIGENT REFRIGERATOR” AND “artificial intelligence”	4	135	139
“INTELLIGENT REFRIGERATOR” AND “internet of things”	8	155	163
Total	497	5303	5800

This systematic review focuses on papers published from 2000 when LG Electronics invented R-S73CT the first smart refrigerator, and research on technology involvement in refrigerators began to increase [23,91]. Combination of keywords were used for the search, as follows: “Smart refrigerator”, “Smart refrigerator AND “internet of things”, “Smart refrigerator” AND “artificial intelligence”, “Smart Fridge”, “Grocery Ordering Systems”, “Fridge” AND “IoT” AND “AI” AND “Smart refrigerator”, “smart food storage”, “Intelligent Refrigerator”, “Intelligent Refrigerator” AND “artificial intelligence”, “Intelligent Refrigerator” AND “internet of things”. The search conducted in this paper uses ProQuest and Google Scholar databases. Only papers written in English and published in journals and conferences are selected for this paper. Publications were excluded if they had or were: 1) access restrictions for the entire paper, 2) papers for class projects, 3) proposed systems that applied IoT or AI on other than home food storage compartments, 4) relevant but the mechanism of household interaction was missing, and 5) newspapers, wire feeds, blogs, podcasts, websites, patents, dissertations and theses, books, and magazines are excluded. Table 1 shows a summary of the search results using the keywords and search engines.

The initial search resulted in 5800 possible papers to review, as shown in Fig. 2. Before the initial screening 226 duplicated results were removed. From the initial screening of the titles and abstracts of each result and applying the exclusion criteria, 72 papers remained in this systematic review. There were 42 conference papers and 28 published papers. Two papers were excluded after the full-text reading for reasons number 4 and 5.

Here are some examples of excluded papers after the screening of the title and the abstract with exclusion explanations:

- A paper titled “Design of High-Efficiency Refrigerator Test System for Industrial Internet of Things” by Xian et al. [90] was excluded as it focuses on reducing the refrigerator’s power consumption.
- A paper titled “A Taxonomy and Survey of IoT Cloud Applications” by Pflanzner et al. [68] It is about the application of IoT in general. It talks about the smart home appliance and the smart refrigerator as an example, stating the product of Samsung and its features. No details were given, and no solution was proposed.
- A paper titled “Application of Affordance Factors for User-Centered Smart Homes: A Case Study Approach” by Younjoo et al. [15] is a case study about a smart home’s central interface that helps the user to view and manage home appliances, the smart refrigerator was one of the examples, and no details were given to support this systematic review.
- A paper titled “Older Adult Segmentation According to Residentially-Based Lifestyles and Analysis of Their Needs for Smart Home Functions” by Jiyeon et al. [92] is about old people’s lifestyles and how vital smart homes are to them. Also, it brought the smart refrigerator as an example of what it could have as features briefly.
- A paper titled “Monitoring in IOT enabled devices” by Gupta. [29] It talks about letting the smart refrigerator adjust the temperature itself based on the weather to reduce power consumption. It has nothing related to tracking food items inside the storage compartment.

- A paper titled “Intrusion Detection In Internet Of Things (IOT)” by Anthony et al. [64] It talks about the security of IoT and provides a block diagram for how smart refrigerators are being connected through Bluetooth to a smartphone and showed it as an example with a weighing sensor. However, it is still about how hackers can easily interrupt it.
- A paper titled “Safety of Food and Food Warehouse Using VIBHISHAN” by Khan et al. [43] talks about food safety and gives some ideas and tools to monitor it inside warehouses. There was no module or test implemented or even conceptual solution to support inclusion.
- A paper titled “Multi-Class Fruit Classification Using Efficient Object Detection and Recognition Techniques” by Khan and Debnath [42] This paper is good as it helps in the fruit recognition method using AI. But it is being excluded as its focus is only on fruit recognition and removing the noise of the pictures to make a proper decision on the taken images. Therefore, it does not include the mechanism of tracking and reporting food items inside the refrigerator to the household, which is the focus of this work. Maybe later, this paper can be used to understand how to test the AI approach and get results of the developed unified framework.

Here are two examples of why papers were excluded after full-text reading:

- The paper titled “Next Generation Smart Fridge System using IoT” by Bhatt et al. [10] is closely relevant to this review but was excluded. It was a class project.
- The paper is titled “Inventory Management of the Refrigerator’s Produce Bins Using Classification Algorithms and Hand Analysis.” by Morris et al. [57] was excluded as it focuses only on hand detection and allowing the recognition model to clear the picture using a CNN classifier from the background and keeps the food items’ image for comparison. There is no mention of how the household interacted with the system. Maybe later, this paper can help to use the AI approach better and get robust results from the developed unified framework as an extension of this systematic review.

### 3. Household replenishment systems analysis

This section analyzes the selected papers based on the used approach and the implemented components. The 70 reviewed papers apply to the IoT technology or a combination of IoT and AI technologies, also referred to as Artificial Intelligence of Things (AIoT), and how those technologies can work with perishable item storage compartment systems.

Twenty-seven papers shed more light on the combination of AIoT components with perishable items storage compartments. Ten showed the results based on conceptual assumptions, 15 based on simulations, and 2 on the implementation of the system.

Forty-three papers drew the connection between IoT components and perishable item compartments. Twenty showed the solutions with conceptually assumed results, 17 of them were based on simulation, and six were on implementation of the system.

This section contains subsections that detail the IoT and AI components used in each reviewed paper. Each subsection will end with a table that summarizes the discussed matter. The component model and type will be written as mentioned in the reviewed papers. A checkmark (X) will be against the authors’ name if the type or model of the used component is included but not specified. Otherwise, it will be left blank or not listed in the subsections’ tables as it was not used.

This section ends with highlighted contributions, and limitations of and observations from each reviewed paper in Table 20 and Table 21, respectively. A complete table summarizes the subsections of the used components in Table 22. It forms the base for a data mining technique and a decision tree for a continuous variable decision. The frequency of used components among the reviewed papers is used to define the relationship between the used components and form them into categories. The continuous variable decision is used for the variables that depend on the outcome of each other, which applies to this paper. Therefore, this type of decision tree method is used, as shown in Table 23.

Table 2  
Articles in which the authors used IR/Ultrasonic sensors

Article	Year	IR	Ultrasonic
Loh et al. [50]	2004	X	
Nayak et al. [62]	2011	X	
Lloret et al. [49]	2012		X
Sandholm et al. [75]	2014	X	
Prapulla et al. [69]	2015	X	
Panchal et al. [67]	2015	X	X
Edward et al. [19]	2017	X	
Shama et al. [77]	2017	X	
Wu et al. [89]	2017	X-Distance (Sharp GP2Y0A41SKOF)	
Anand et al. [6]	2018		HC-SR04T
Hossain et al. [34]	2018		HC-SR04T
Barfeh et al. [8]	2019	X	
Bayya [9]	2019	X	
Khan et al. [41]	2019	X	
Shariff et al. [78]	2019	X	X
Sharma et al. [79]	2019	X	
Mallikarjun et al. [52]	2020	X	HC-SR04T
Velasco et al. [86]	2020		X
Das et al. [16]	2021	X	X
Jaipriya et al. [37]	2021		X
Krishnamoorthy et al. [46]	2021		X
Nejekar et al. [63]	2022	X	

### 3.1. IR/ ultrasonic sensors

These sensors measure the liquid level inside items stored and the distance between them in the storage compartment. IR the infrared radiation sensors work by sending several light lines as signals across the compartment. Then, a measurement will be sent to the controller when a new or moved object interrupts the light to measure the distance or the remaining quantity level [89]. Ultrasonics are sound sensors that send sound signals and based on how the objects reflect, the distance will be measured [6]. A connected database will be updated with the feedback captured from these sensors, where the items' status will be calculated accordingly [83]. Some authors used these sensors, as shown in Table 2.

### 3.2. Climate sensors

This component is different than the default built-in climate sensor. Temperature and humidity sensors are placed inside the storage compartment. These sensors give feedback to a connected device (controller or household mobile device) in the network or update the connected database with an up-to-date status of the storage climate. Some climate sensors can be programmed to control the climate inside the storage compartment from a distance. Several types of climate sensors are included in some of the reviewed papers, as shown in Table 3.

### 3.3. Light sensors

A light sensor has different uses. Sometimes it can check the light inside the storage compartment. Once it senses light, it can trigger the connected equipment to function or send an alert that the refrigerator door is open [89]. A list of authors used this type of sensor in their systems, as shown in Table 4.

Table 3  
Articles in which the authors used climate sensors

Article	Year	Temperature/Humidity sensors
Lloret et al. [49]	2012	X
Gürüler [30]	2015	X-NTC negative temperature coefficient
Osisanwo et al. [65]	2015	X
Edward et al. [19]	2017	DS18B20
Qiao et al. [70]	2017	DS18B20
Shama et al. [77]	2017	LM35
Nasir et al. [61]	2018	DHT11
Zhongmin et al. [93]	2018	Temperature (DS18B20) /Humidity (DHT11)
Ahmed and Rajesh [4]	2019	Temperature (LM35) /Humidity (DH22)
Shariff et al. [78]	2019	X
Dong et al. [18]	2020	DHT11
Chakilam et al. [13]	2021	DHT11
Gupta et al. [28]	2021	Arduino Uno (Steinhart)
Jain et al. [36]	2021	DHT11
Krishnamoorthy et al. [46]	2021	X
Nadar et al. [58]	2021	DHT11
Sane et al. [76]	2021	Temperature (LM35) /Humidity (DH11)
Nejekar et al. [63]	2022	DHT11

Table 4  
Articles in which the authors used light sensors

Article	Year	Light sensor
Lloret et al. [49]	2012	X
Wu et al. [89]	2017	X
Dong et al. [18]	2020	GY-320 to trigger the camera
Chakilam et al. [13]	2021	LDR to trigger the camera

Table 5  
Articles in which the authors used gas sensors

Article	Year	Gas sensor
Lloret et al. [49]	2012	X
Anand et al. [6]	2018	ME3
Nasir et al. [61]	2018	MQ3
Zhongmin et al. [93]	2018	ZP07
Ringe et al. [72]	2019	X
Shariff et al. [78]	2019	X
Chakilam et al. [13]	2021	MQ3
Gull et al. [27]	2021	eNose System (Fruits/Vegetables: MQ3, Meat:MQ135)
Jain et al. [36]	2021	MQ3
Nejekar et al. [63]	2022	MQ series

### 3.4. Gas sensors

A gas sensor is designed to sense and measure the gas generated by food items such as vegetables, fruits, and meat to identify the item type and predict spoilage. Several types of gas sensors were involved in some of the proposed systems, as shown in Table 5.

Table 6  
Articles in which the authors used door open-close sensors

Article	Year	Door open-close sensor
Loh et al. [50]	2004	Switch
Nayak et al. [62]	2011	X
Kaldeli et al. [38]	2013	X
Lloret et al. [49]	2012	X
Sandholm et al. [75]	2012	LED
Son et al. [82]	2014	NFC
Kale et al. [39]	2015	X
Kwon et al. [47]	2016	Switch
Esmaili [5]	2017	X
Hafidh et al. [32]	2017	X
Hossain et al. [34]	2018	X
Khan et al. [41]	2019	Camera
Sharma et al. [79]	2019	X
Mohammad et al. [56]	2020	X
Velasco et al. [86]	2020	X
Datey [17]	2021	X
Krishnamoorthy et al. [46]	2021	Servo motor
Lee et al. [48]	2021	X
Sharma et al. [80]	2021	X

### 3.5. Door open-close sensors

A door open-close sensor is a sensor that is used to trigger the connected equipment whenever the storage compartment's door is opened and closed or remains open or alerts the household that the door is open. Some of the reviewed papers used several types of these sensors, as shown in Table 6.

### 3.6. Weight sensors

Weight sensors can be placed or connected to the bottom of the storage shelves. Then these sensors give feedback to a connected device in the network or update the connected database with an up-to-date status of stored items' weight as a quantity measurement. Some of the reviewed papers used several types of these sensors in their systems, as shown in Table 7.

### 3.7. GSM module

The Global System for Mobile is a device that has a SIM port that enables the linked device to have telecommunication ability. It enables the storage compartment to send alerts based on the received data from the controller and sensors. Also, it allows the household to control the refrigerator by sending commands via SMS [50]. Some authors applied a GSM module, as shown in Table 8.

### 3.8. RFID system

The RFID system identifies objects from their identity tags. It consists of two parts the RFID tag and the RFID reader. The RFID tag is where the information about the thing is stored. It has two types, and each type has several models that read the associated tags [44]. One requires a long-life battery, and the information recorded on it can be modified and updated at any time, and it is called an RFID active tag. The other tag is called an RFID passive tag, and this type of tag does not require a battery as the tag will automatically activate once it becomes close to the RFID reader, and the information recorded in this tag cannot be changed. A tag will be attached to each object,

Table 7  
Articles in which the authors used weight sensors

Article	Year	Weight sensors
Lloret et al. [49]	2012	X
Kale et al. [39]	2015	X
Prapulla et al. [69]	2015	X
Goeddel et al. [26]	2017	X
Hafidh et al. [32]	2017	X
Qiao et al. [70]	2017	X
Anand et al. [6]	2018	HX711
Fujiwara et al. [24]	2018	KD-320 by TANITA and HX711-M
Hossain et al. [34]	2018	Flexi Forse A401
Rezwan et al. [71]	2018	X
Zhongmin et al. [93]	2018	X
Ahmed and Rajesh [4]	2019	X
Bayya [9]	2019	X
Khan et al. [41]	2019	HX711
Narayan et al. [60]	2019	HX711
Ringe et al. [72]	2019	X
Kore et al. [45]	2020	HX711
Mallikarjun et al. [52]	2020	HX711
Velasco et al. [86]	2020	X
Datey [17]	2021	X
Gull et al. [27]	2021	HX711
Gupta et al. [28]	2021	HX711
Jain et al. [36]	2021	HX711
Nadar et al. [58]	2021	HX711
Sane et al. [76]	2021	X
Nejekar et al. [63]	2022	HX711

Table 8  
Articles in which the authors used GSM module

Article	Year	GSM module
Loh et al. [50]	2004	X
Nayak et al. [62]	2011	X
Rouillard [73]	2012	X
Kaldeli et al. [38]	2013	X
Gürüler [30]	2015	X
Kale et al. [39]	2015	X
Panchal et al. [67]	2015	X
Esmaili [5]	2017	X
Hafidh et al. [32]	2017	X
Bayya [9]	2019	X
Shariff et al. [78]	2019	X
Das et al. [16]	2020	X
Datey [17]	2021	X
Jaipriya et al. [37]	2021	X
Krishnamoorthy et al. [46]	2021	X

Table 9  
Articles in which the authors used RFID system

Article	Year	RFID system
Hong et al. [33]	2007	x
Konidala et al. [44]	2011	X
Hou et al. [35]	2013	X
Kaldeli et al. [38]	2013	X
Son et al. [82]	2014	X
Osisanwo et al. [65]	2015	X
Calegari et al. [12]	2016	X
Floarea et al. [23]	2016	X
Hachani et al. [31]	2016	UHF
Esmaili [5]	2017	X
Qiao et al. [70]	2017	X
Shama et al. [77]	2017	X
Anand et al. [6]	2018	X
Hossain et al. [34]	2018	X
Abdel-Basset et al. [2]	2019	X
Bayya [9]	2019	X
Ferrero et al. [22]	2019	X
Shariff et al. [78]	2019	X
Mohammad et al. [56]	2020	X
Jaipriya et al. [37]	2021	X
Nejekar et al. [63]	2022	X

Table 10  
Articles in which the authors used a Barcode system

Article	Year	Barcode system
Luo et al. [51]	2009	X
Rouillard [73]	2012	X
Hou et al. [35]	2013	X
Edward et al. [19]	2017	X
Hossain et al. [34]	2018	X
Abd Elminam et al. [1]	2020	Via Smartphone Camera
Dong et al. [18]	2020	X

and whenever that object becomes close to the RFID reader, it will be activated. The RFID reader will read the information, and then send it to a connected central database. Sometimes auto-scan can be set up to run an overall scan of the stored items to check their availability. The setup can be programmed to be periodically or triggered by another sensor [23]. Some authors applied the RFID system, as shown in Table 9.

### 3.9. Barcode system

The barcode system is a technology that acts similar to the RFID system but it is a manual system. A barcode scanner is required to scan the barcode tag located on any barcoded item. The barcode tag consists of lines and numbers printed in a certain way representing information about an associated item. Then, a scanner reads the information about the scanned item and sends it to a connected database. The information carried by the barcode varies and may contain data such as an item's type, quantity, and expiration date [33,50]. Some papers applied the barcode system, as shown in Table 10.

Table 11  
Articles in which the authors used connection medium

Article	Year	Connection medium
Nayak et al. [62]	2011	Ethernet
Lloret et al. [49]	2012	X
Hou et al. [35]	2013	X
Gürüler [30]	2015	Ethernet
Kale et al. [39]	2015	X
Osisanwo et al. [65]	2015	X
Panchal et al. [67]	2015	X
Kwon et al. [47]	2016	X
Qiao et al. [70]	2017	X
Shama et al. [77]	2017	WiFi
Wu et al. [89]	2017	X
Anand et al. [6]	2018	X
Fujiwara et al. [24]	2018	X
Nasir et al. [61]	2018	X
Rezwan et al. [71]	2018	X
Zhongmin et al. [93]	2018	X
Abdel-Basset et al. [2]	2019	X
Ahmed and Rajesh [4]	2019	X
Barfeh et al. [8]	2019	WiFi
Bayya [9]	2019	X
Narayan et al. [60]	2019	X
Ringe et al. [72]	2019	X
Abd Elminam [1]	2020	Bluetooth
Avinash et al. [7]	2020	X
Kore et al. [45]	2020	X
Mohammad et al. [56]	2020	X
Velasco et al. [86]	2020	WiFi
Chakilam et al. [13]	2021	X
Datey [17]	2021	X
Gull et al. [27]	2021	X
Krishnamoorthy et al. [46]	2021	X
Nadaret al. [58]	2021	X
Sane et al. [76]	2021	X

### 3.10. Connection medium

Most current devices have already been designed to be identified and connected to the internet wirelessly via a WiFi connection or wired via Ethernet and with each other via Bluetooth. Other appliances such as regular refrigerators or storage cabinets require specific IoT equipment to be identified. One of the reviewed papers used an Ethernet connection, a wired network that enables devices to communicate within a local area network and to the internet [30]. Some systems used different types of connections, as shown in Table 11.

### 3.11. Controller

Controllers are used to send and receive information, store data, enable internet access for the connected equipment, and be configured to process the received data. Some of the reviewed papers used an IoT platform, tablets, or personal computers to act as the controller, but with larger data storage. Several types of controllers are installed in the systems conducted by the authors, as shown in Table 12.

Table 12  
Articles in which the authors used controllers in their systems

Article	Year	Controller
Loh et al. [50]	2004	X
Hong et al. [33]	2007	X as a grocery shopping agent
Luo et al. [51]	2009	TOUCHSCREEN
Konidala et al. [44]	2011	PC home server
Nayak et al. [62]	2011	X
Bostanci et al. [11]	2013	X
Kaldeli et al. [38]	2013	PC home server
Son et al. [82]	2014	PC
Gürüler [30]	2015	PIC187J60
Kale et al. [39]	2015	X
Osisanwo et al. [65]	2015	X
Panchal et al. [67]	2015	Arduino Uno
Calegari et al. [12]	2016	Raspberry Pi
Floarea et al. [23]	2016	Arduino SoC model with Intel Edison processor
Kwon et al. [47]	2016	Arduino Uno-Raspberry Pi (Raspbian OS-Linux)
Esmaili [5]	2017	Arduino Uno
Edward et al. [19]	2017	Arduino (NodeMCU) and Raspberry Pi3
Hafidh et al. [32]	2017	Main board
Shama et al. [77]	2017	X
Shweta [81]	2017	X
Wu et al. [89]	2017	Raspberry Pi 2 BV1.1
Anand et al. [6]	2018	Arduino ATmega2560
Hossain et al. [34]	2018	Raspberry Pi
Nasir et al. [61]	2018	Arduino Uno WeMos D1R2 WiFi
Rezwan et al. [71]	2018	Arduino Mega and NodeMCU (ESP8266)
Zhongmin et al. [93]	2018	Aduino Arm Contex – M3 and STM32F103 ARM
Abdel-Basset et al. [2]	2019	X
Ahmed and Rajesh [4]	2019	Arduino Uno
Barfeh et al. [8]	2019	Raspberry Pi
Bayya [9]	2019	Arduino Uno
Khan et al. [41]	2019	Raspberry Pi for Fridge and Node MCU for sensors
Narayan et al. [60]	2019	Arduino Uno R3 AT Mega 328 (IDE Coding) and NodeMCU (ESP8266)
Ringe et al. [72]	2019	Raspberry Pi
Shariff et al. [78]	2019	Renesas GR Peach with (RFID and GPS) built-in: but not utilized
Sharma et al. [79]	2019	X
Abd Elminam et al. [1]	2020	Arduino Mega
Avinash et al. [7]	2020	X
Das et al. [16]	2020	Arduino Uno Atmega
Dong et al. [18]	2020	Arduino Leonardo

### 3.12. Internet protocol

The Internet Protocol is a set of regulations controlling network communication with a given (IP address) that identifies machines connected to the internet or locally. An IP address will act as a unique identifier assigned to a smart refrigerator to allow the household to monitor and control that particular refrigerator [44]. Several types of internet protocols are applied in reviewed papers, as shown in Table 13.

Table 12  
(Continued)

Article	Year	Controller
Kore et al. [45]	2020	Raspberry Pi B3
Mallikarjun et al. [52]	2020	Raspberry Pi B3
Mohammad et al. [56]	2020	Arduino
Saha et al. [74]	2020	Raspberry Pi B3 – Python coding and OpenCV library for images
Chakilam et al. [13]	2021	NodeMCU-8266 and Raspberry Pi
Datey [17]	2021	X
Gull et al. [27]	2021	Arduino Uno IDE
Gupta et al. [28]	2021	Arduino Uno
Jain et al. [36]	2021	Arduino and Raspberry Pi 3B+
Jaipriya et al. [37]	2021	X
Krishnamoorthy et al. [46]	2021	Raspberry Pi
Nadar et al. [58]	2021	NodeMcu
Sane et al. [76]	2021	Arduino Uno ATmega
Sharma et al. [80]	2021	Raspberry Pi
Nejekar et al. [63]	2022	Raspberry Pi

Table 13  
Articles in which the authors applied an Internet Protocol

Article	Year	Internet protocol
Konidala et al. [44]	2011	HTTPS
Nayak et al. [62]	2011	X
Lloret et al. [49]	2012	X
Sandholm et al. [75]	2014	support HTTP POST, GET, JSON
Osisanwo et al. [65]	2015	X
Kwon et al. [47]	2016	HTTP
Hossain et al. [34]	2018	MQTT
Abdel-Basset et al. [2]	2019	X
Narayan et al. [60]	2019	MQTT
Ringe et al. [72]	2019	MQTT and COAP: to support images
Mallikarjun et al. [52]	2020	MQTT

### 3.13. IoT platform

An IoT Platform allows objects to communicate, exchange, and store information via communication channels, databases, and internet protocols. According to Floarea et al. [23] there are four types of IoT platforms: 1) Machine-to-machine connectivity (M2M), which handles the communication between the IoT-connected components via a telecommunication network, but cannot process data; 2) Infrastructure as a Service (IaaS) acts as a backend server over the internet, allowing individuals to have a space with full access to control, store, and process data (platform is compatible with many operating systems); 3) Hardware-Specific software is exclusive software that operates devices; and 4) Consumer/Enterprise software extensions generally come as packages of multi-functional software programs and act as an IoT platform [3,23].

To operate an IoT platform, several features must be included [3,23]: 1) connectivity and normalization for the data flow assurance and accuracy, 2) device management where the connected devices are managed appropriately, 3) a scalable database that can accommodate vast amounts of data, 4) managing data from connected devices to take appropriate actions, 5) the ability to generate analytics reports based on individual preferences, 6) a dashboard to allow individuals to view meaningful information, 7) additional tools that allow testing, implementing, and model-

Table 14  
Articles in which the authors utilized IoT Platform

Article	Year	IoT platform
Rouillard [73]	2012	Database for pricing the products using Pricing
Hou et al. [35]	2013	Cloud server to process data and prepare a shopping list
Sandholm et al. [75]	2014	X
Osisanwo et al. [65]	2015	X
Floarea et al. [23]	2016	Google Cloud
Hachani et al. [31]	2016	Cloud service
Kwon et al. [47]	2016	Apache Web Server-MSQL
Goeddel et al. [26]	2017	Cloud
Qiao et al. [70]	2017	Cloud-based Platform
Shama et al. [77]	2017	X
Wu et al. [89]	2017	Google Firebase
Anand et al. [6]	2018	Google Firebase
Hossain et al. [34]	2018	ThingSpeak
Nasir et al. [61]	2018	ThingSpeak
Rezwan et al. [71]	2018	Web Application GUI
Ahmed and Rajesh [4]	2019	Google Firebase
Ferrero et al. [22]	2019	Google Firebase
Khan et al. [41]	2019	Google Firebase
Ringe et al. [72]	2019	X
Dong et al. [18]	2020	Ubidots dashboard and database
Kore et al. [45]	2020	API
Mallikarjun et al. [52]	2020	Google Firebase
Nagarajua et al. [59]	2020	API
Velasco et al. [86]	2020	Cloud Server
Chakilam et al. [13]	2021	Google AI and Google Cloud
Che Soh et al. [14]	2021	Ubidots dashboard and database
Jain et al. [36]	2021	Google Firebase
Nadar et al. [58]	2021	X
Sane et al. [76]	2021	X

ing, and 8) an external interface that allows the IoT platform to be expandable and to be monitored from a mobile device. Different IoT platforms use different systems, as shown in Table 14.

### 3.14. Tablet/Touchscreen/PC

Smart refrigerators can be connected to household devices and the internet through external tablets or personal computers; some use built-in touchscreens. With a pre-installed application or web application, this tablet could communicate with all devices connected to the network. It also can receive, process, store, update and send information to a central database or a household mobile device. PC acts like tablets with more capabilities; they usually have extra capacity, provide a convenient programming environment, and serve as a home server. While touchscreens vary, some are like tablets and others just for a few functions. Here is a list of authors using different types of these devices in their systems, as shown in Table 15.

### 3.15. Mobile application

Specifically designed software allows mobile devices to interact with the storage compartment in many ways, such as retrieving information, monitoring, controlling, and approving shopping lists. Several types of mobile applications are used in different papers, as shown in Table 16.

Table 15  
Articles in which the authors utilized Tablets/PC in various systems

Article	Year	Tablet/Touchscreen/PC
Hong et al. [33]	2007	PC
Luo et al. [51]	2009	TOUCHSCREEN
Nayak et al. [62]	2011	PC
Lloret et al. [49]	2012	PC
Bostanci et al. [11]	2013	PC application
Kaldeli et al. [38]	2013	PC
Hou et al. [35]	2013	Screen for display only
Son et al. [82]	2014	PC
Prapulla et al. [69]	2015	PC
Hachani et al. [31]	2016	Touchscreen with Voice message alerts
Esmaili [5]	2017	X
Shama et al. [77]	2017	X
Wu et al. [89]	2017	TOUCHSCREEN
Fujiwara et al. [24]	2018	TOUCHSCREEN
Pachón et al. [66]	2018	User Interface
Zhongmin et al. [93]	2018	TOUCHSCREEN
Shariff et al. [78]	2019	PC
Avinash et al. [7]	2020	X
Chakilam et al. [13]	2021	Dashboard Cayenne
Datey [17]	2021	X
Gull et al. [27]	2021	PC
Jaipriya et al. [37]	2021	X
Krishnamoorthy et al. [46]	2021	X
Sane et al. [76]	2021	X

### 3.16. Offline-database

In the reviewed papers, databases store information captured or received from the connected devices within the local network so the household can manage it. Different types of databases were used in the reviewed papers, as shown in Table 17.

### 3.17. Webcam/camera module

Webcam means the reviewed paper used a camera to capture low-resolution images. A camera module means that the reviewed paper used a camera connected with a programmed device for image recognition and processing and/or higher resolution images. Some Authors used different types of cameras, as shown in Table 18.

### 3.18. Recognition module

The recognition module in this paper refers to machine learning or deep learning to recognize images, voices, and captured data. Algorithms and models enable the system to recognize and make decisions on the stored items' status [11,55]. Also, they can use facial recognition to learn the consumption habits of each household member and form patterns accordingly, patterns of food use [11]. several methods of recognition modules are applied, as shown in Table 19.

Table 16  
Articles in which the authors used Mobile Application

Article	Year	Mobile application
Lloret et al. [49]	2012	X
Bostanci et al. [11]	2013	X
Hou et al. [35]	2013	As a User Interface for the manual entry of items
Prapulla et al. [69]	2015	Email
Panchal et al. [67]	2015	Android application
Calegari et al. [12]	2016	X
Hachani et al. [31]	2016	X
Kwon et al. [47]	2016	X
Edward et al. [19]	2017	X
Qiao et al. [70]	2017	Intelligent terminal
Shama et al [77]	2017	X
Wu et al. [89]	2017	X
Fujiwara et al. [24]	2018	X
Hossain et al. [34]	2018	C
Nasir et al. [61]	2018	PushBullet
Rezwan et al. [71]	2018	X
Ahmed and Rajesh [4]	2019	As User Interface
Barfeh et al [8]	2019	Android
Ferrero et al. [22]	2019	For Voice interaction using Google Assistant SDK
Narayan et al. [60]	2019	X
Ringe et al. [72]	2019	Android
Abd Elminam et al. [1]	2020	X
Avinash et al. [7]	2020	X
Das et al. [16]	2020	X
Kore et al. [45]	2020	X
Mallikarjun et al. [52]	2020	X
Saha et al. [74]	2020	X
Velasco et al. [86]	2020	Android application
Datey [17]	2021	X
Gupta et al. [28]	2021	X
Jain et al. [36]	2021	Used for monitoring and shopping
Krishnamoorthy et al. [46]	2021	X
Nadar et al. [58]	2021	X
Sane et al. [76]	2021	X
Sharma et al. [80]	2021	X
Nejekar et al. [63]	2022	X

### 3.19. Approach and contribution of reviewed papers

The approach each reviewed paper used on the introduced systems towards a smart refrigerator is shown in Table 20, along with the papers' contribution regarding storage compartments. Table 21 shows the limitations of and observations from each reviewed paper.

### 3.20. Analysis summary of the utilized components

This subsection provides a table that summarizes each reviewed paper's used components. It consists of a list of authors in chronological order, and when a tie occurs between years, the authors are listed alphabetically. A checkmark (X) will indicate that the system introduced by each author included that component. The checkmark will be

Table 17  
Articles in which the authors used systems with an offline-database

Article	Year	Offline-database
Luo et al. [51]	2009	X
Lloret et al. [49]	2012	X
Panchal et al. [67]	2015	X
Goeddel et al. [26]	2017	X
Shweta [81]	2017	X
Nasir et al. [61]	2018	PLX-DAQ and MS Excel
Pachón et al. [66]	2018	X
Barfeh et al. [8]	2019	X
Sharma et al. [79]	2019	X
Das et al. [16]	2020	X
Kore et al. [45]	2020	A database uses the root (.CVS)
Datey [17]	2021	X
Krishnamoorthy et al. [46]	2021	X
Sharma et al. [80]	2021	X

Table 18  
Types of cameras used in the literature

Article	Year	Webcam/camera module
Bostanci et al. [11]	2013	Camera module
Kwon et al. [47]	2016	Camera module
Goeddel et al. [26]	2017	Camera module with Python coding
Shweta [81]	2017	Camera module
Wu et al. [89]	2017	X
Anand et al. [6]	2018	Adafruit
Pachón et al. [66]	2018	Camera module
Khan et al. [41]	2019	Camera module
Ringe et al. [72]	2019	Webcam
Sharma et al. [79]	2019	Camera module
Avinash et al. [7]	2020	Camera module
Dong et al. [18]	2020	Camera module
Kore et al. [45]	2020	Camera module
Mallikarjun et al. [52]	2020	Camera module-INTEXT-305EC
Mohammad et al. [56]	2020	Camera module-CMOS
Nagarajua et al. [59]	2020	Camera module
Saha et al. [74]	2020	Camera module
Velasco et al. [86]	2020	Webcam
Chakilam et al. [13]	2021	Camera module
Che Soh et al. [14]	2021	Camera module
Datey [17]	2021	Camera module
Jain et al. [36]	2021	Camera module
Lee et al. [48]	2021	Camera module
Sane et al. [76]	2021	Camera module
Sharma et al. [80]	2021	Camera module
Nejekar et al. [63]	2022	Webcam

Table 19  
The recognition modules used in the literature

Article	Year	Recognition module
Bostanci et al. [11]	2013	Fuzzy logic algorithm
Sandholm et al. [75]	2014	Google Image search engine
Kwon et al. [47]	2016	Fisher's Linear Discriminant Analysis algorithm
Goeddel et al. [26]	2017	X
Shweta [81]	2017	Aging algorithm-Machine Learning
Anand et al. [6]	2018	X
Fujiwara et al. [24]	2018	X
Hossain et al. [34]	2018	X
Pachón et al. [66]	2018	CNN-to train the data for image recognition
Gao et al. [25]	2019	Deep Learning SSD algorithm
Khan et al. [41]	2019	X
Sharma et al. [79]	2019	CNN and DNN
Avinash et al. [7]	2020	X
Dong et al. [18]	2020	CNN-to train the data for image recognition and Deep Learning Framework Caffe
Kore et al. [45]	2020	Machine Learning ImageNet Classifier algorithm
Mallikarjun et al. [52]	2020	Machine Learning K-means Classifier algorithm
Mohammad et al. [56]	2020	CNN, Transfer Learning Technique and Inception V3 pre-trained model
Nagarajua et al. [59]	2020	CNN-to train the data for image recognition
Saha et al. [74]	2020	Machine Learning YOLO V3, Tiny YOLO and ImageAI library for training data
Chakilam et al. [13]	2021	CNN YOLO
Che Soh et al. [14]	2021	Faster R-CNN and SSD Mobilenet for object detection
Datey [17]	2021	X
Gull et al. [27]	2021	Machine learning to make a decision based on data from sensors
Jain et al. [36]	2021	CNN-Inception-V3
Lee et al. [48]	2021	CNN and Object segmentation and argumentation Deep learning
Sane et al. [76]	2021	X
Sharma et al. [80]	2021	Transfer Flow Object classifier Deep Learning for image recognition and SMO Self Organizing Map for user behavior (NN)

entered differently according to the footnotes for the recognition module and implementation type columns. The footnotes are repeated before each page break in Table 22.

Table 22 presents all reviewed papers in chronological order. The chronological order and the check marks indications show the revolution of the household replenishment systems over the years. Therefore, it will be easy for the researchers or manufacturers to keep the trendy components or eliminate them as they become outdated, based on their objective. The researchers and manufacturers could also consider other factors shown in other tables to decide on the targeted combination. For example, Table 23 and Table 24 could be used as guidelines to assign weight coefficients among the presented components based on their trendiness, frequency, functionalities, etc.

### 3.21. Frequency of component selection using IoT and AIoT approaches

This subsection reports the frequency of component selection based on the used approach. Using MS Excel to find the frequency between the used components and approaches Table 23 shows the reviewed papers with different approaches in using IoT and AIoT components toward home perishable items storage compartments. The following are charts captured from the performed analysis on these used components based on the frequency result in Table 23.

Of the 70 reviewed papers, 43 used IoT technology to convert perishable items storage compartments into smart devices. Twenty-seven papers used AIoT technology. The camera and recognition modules were not used in the IoT approach, while all AIoT systems use the recognition module. That is because the IoT approach does not mimic human interaction to make decisions, but AIoT does. Moreover, the GSM module was only used once by a system

Table 20  
Approach and contribution of reviewed articles

Article	Year	Approach	Scientific/practical contributions
Loh et al. [50]	2004	IoT	Proposed a system uses IR Sensors for empty space sensing.
Hong et al. [33]	2007	IoT	Proposed a mathematical model for optimal replenishment policies using an RFID system.
Luo et al. [51]	2009	IoT	A novel smart refrigerator database to keep track of the nutrition of stored items based on a barcode reader using Microsoft SQL.
Konidala et al. [44]	2011	IoT	Security framework for RFID-based applications based on cryptographic methods and primitives and proposed system that keeps track of newly added items.
Nayak et al. [62]	2011	IoT	Proposed an IR sensing system that monitors a stored item's stock level and generates an auto order to the nearest store.
Lloret et al. [49]	2012	IoT	Proposed a system that enables the refrigerator to inform the household via a Twitter account about the stored items' levels.
Rouillard [73]	2012	IoT	Proposed a system to alert the household about stored items' levels using a smartphone for barcode scanning, voice recognition, instant messaging, and an RFID scanner.
Bostanci et al. [11]	2013	AIoT	Uses fuzzy logic and neural network Hopfield NN single layer, One for food and one for facial recognition.
Hou et al. [35]	2013	IoT	Proposed a food management system using barcodes, RFID scanners, and manual entry of non-tagged items.
Kaldeli et al. [38]	2013	IoT	Proposed an RFID system to identify and monitor the stored items and a GSM module for communication.
Sandholm et al. [75]	2014	AIoT	Introduced a novel system that used a Google Image search engine for stored items recognition.
Son et al. [82]	2014	IoT	Proposed a system for diet management for wellness service refrigerators that allows the user to track stored and consumed RFID-tagged items.
Gürüler [30]	2015	IoT	Proposed a GSM module to enable a user to communicate with a refrigerator via SMS to learn the system status.
Kale et al. [39]	2015	IoT	Used a Contextual Inquiry (CI) for a better understanding of the household behavior in storing food items to help in designing designated trays for specific food items for tracking purposes.
Osisanwo et al. [65]	2015	IoT	Discussed the benefits and security challenges of smart refrigerators.
Prapulla et al. [69]	2015	IoT	Introduced a system that tracks stored items and communicates with the user via a GSM module and suggested access to be given to stores.
Panchal et al. [67]	2015	IoT	Introduced a system that tracks stored items and notifies the user via a GSM module for placing an order.
Calegari et al. [12]	2016	IoT	Proposed a tracking system for stored items using RFID and to be connected to other home appliances.
Floarea et al. [23]	2016	IoT	Introduced a system that tracks stored food items using RFID and provides a detailed description of the IoT platform.
Hachani et al. [31]	2016	IoT	Proposed a system that tracks the stored items using RFID and allows the refrigerator to alert the user via a voice messaging system embedded in an attached tablet.
Kwon et al. [47]	2016	AIoT	Introduced a system for stored items' recognition using Fisher's Linear Discriminant Analysis as a classifying algorithm.
Esmaeili [5]	2017	IoT	Proposed an MS Visual Studio application to send email alerts to users about the stored items' status captured by RFID scanners. Provides some input components prices.
Edward et al. [19]	2017	IoT	Proposed a system with JavaScript code used for auto-replenishment using an IR sensor and barcode scanner.
Goeddel et al. [26]	2017	AIoT	Proposed a system with algorithms for shopping list creation using image recognition and weighing sensors.
Hafidh et al. [32]	2017	IoT	Proposed a system that tracks the stored items quantity and uses email to alert a user.
Qiao et al. [70]	2017	IoT	Proposed an automated system that tracks stored items and alerts a user about the system status using an RFID scanner and weighing sensors.

Table 20  
(Continued.)

Article	Year	Approach	Scientific/practical contributions
Shama et al. [77]	2017	IoT	Proposed a system that allows users to remotely adjust the refrigerator temperature and know stored item levels by using RFID.
Shweta [81]	2017	AIoT	Proposed a system embedded with an Aging algorithm to identify the stored vegetable ages. A voice message is used to alert the user.
Wu et al. [89]	2017	IoT	Proposed an application that enables the user to receive a captured image each time the refrigerator is opened and manually update the stored item's status.
Anand et al. [6]	2018	AIoT	Proposed a system that tracks the stored items' status. Food quality could be known via its expiration date and predicted using gas sensors. The user could be updated via a mobile application.
Fujiwara et al. [24]	2018	AIoT	Proposed a system using speech recognition – for hands-free updates- where the user could identify items by names in a synchronous or asynchronous order. The system uses a weight sensor for each shelf.
Hossain et al. [34]	2018	AIoT	Proposed a conceptual framework for a neighborhood fridge network that allows tracking stored items in multiple connected refrigerators.
Nasir et al. [61]	2018	IoT	Proposed a system that measures the food quality using gas sensors and alerts a user via SMS, mobile application, or email.
Pachón et al. [66]	2018	AIoT	Proposed a system that detects produce using a region-based CNN tool.
Rezwan et al. [71]	2018	IoT	Proposed a prototype for a storage compartment that could measure the stored items' quantity and alert the user at reorder point. Components prices for installation are included.
Zhongmin et al. [93]	2018	IoT	Proposed a system that tracks the food quantity and quality using weight and gas sensors. Describes selected components. A touchscreen is embedded to allow user interaction.
Abdel-Basset et al. [2]	2019	IoT	Proposed an IoT system for a Decision Support System on best food selection using an RFID scanner.
Ahmed and Rajesh [4]	2019	IoT	Proposed a system that tracks the food quantity using a weight sensor and alerts the user via a mobile application.
Barfeh et al. [8]	2019	IoT	Proposed a configuration of a controller to label each weight sensor with the item name for replenishment purposes.
Bayya [9]	2019	IoT	Proposed a system that tracks stored items' quantity using a weight sensor and RFID scanner. It alerts the user via a GSM module every two hours.
Ferrero et al. [22]	2019	IoT	Proposed a system that tracks stored items and their expiration using RFID. Reviewed some scientific contributions regarding smart refrigerators.
Gao et al. [25]	2019	AIoT	Proposed a system for stored items' recognition using YOLO and SSD framework.
Khan et al. [41]	2019	AIoT	Proposed a system using CNN to measure the number of eggs. Auto online ordering system.
Narayan et al. [60]	2019	IoT	Proposed a prototype framework of an inventory tracking system based on items weight like Amazon Dash Smart Shelf.
Ringe et al. [72]	2019	IoT	Introducing COAP network protocol for camera connection to the cloud to track stored items' status.
Shariff et al. [78]	2019	IoT	Proposed a system that tracks the stored item status and uses a fire sensor detector MQ6 for safety.
Sharma et al. [79]	2019	AIoT	Proposed a system with CNN and DNN for image recognition and identification, as well as an IR camera for thermal images.

Table 20  
(Continued.)

Article	Year	Approach	Scientific/practical contributions
Abd Elminam et al. [1]	2020	IoT	Proposed a system that uses an iPhone camera to scan items' barcodes and alert the user of the expiration date of items. A temperature sensor turns on the fan to adjust the refrigerator's temperature.
Avinash et al. [7]	2020	AIoT	Proposed a system with CNN for image recognition and identification, as well as making images clearer by applying CNN algorithms.
Das et al. [16]	2020	IoT	Proposed a system that enables the household to get notifications via SMS and mobile application for items status.
Dong et al. [18]	2020	AIoT	Proposed a system with a camera model using CNN, barcode and OCR (optical character recognition) for object recognition of stored items.
Kore et al. [45]	2020	AIoT	Proposed a system with a machine learning approach for image recognition and weighing sensors for stored items' quantity. Twilio messaging system is used for SMS alerts.
Mallikarjun et al. [52]	2020	AIoT	Proposed a system with a K-means machine-learning classifier to classify stored vegetables along with ultrasonic, IR, and weight sensors for items quantity.
Mohammad et al. [56]	2020	AIoT	Proposed an Inception V3 pre-trained model for stored item's detection along with an RFID scanner for item recognition. The components' price list included.
Nagarajua et al. [59]	2020	AIoT	Proposed a system with CNN and NLP tools to recognize household voice commands and stored items recognition respectively.
Saha et al. [74]	2020	AIoT	Proposed a system with machine learning tools for items recognition. Compares YOLO and COCO detection models.
Velasco et al. [86]	2020	IoT	Proposed a system that tracks the stored items' status and used the Temboo website to send data to Dropbox where the household can check on the refrigerator status.
Chakilam et al. [13]	2021	AIoT	Proposed a system that tracks the stored items' status with the YOLO approach for item recognition.
Che Soh et al. [14]	2021	AIoT	Proposed a system that identifies and tracks the stored items' status with a deep learning approach. Provided a comparison between SSD and R-CNN models. Uses Telegram to alert the household.
Datey [17]	2021	AIoT	Proposed a system that uses CNN and Regression prediction algorithms to suggest seasonal fruits and vegetables to the household.
Gull et al. [27]	2021	AIoT	Proposed an e-nose system for food item recognition using a machine learning approach (Decision Tree Advanced ID3 model) along with a weight sensor for quantity tracking.
Gupta et al. [28]	2021	IoT	Proposed a system that tracks stored items via weighing sensors and allows the user to control the storage climate using a mobile application.
Jain et al. [36]	2021	AIoT	Proposed a system that used weight sensors for items quantity, a gas sensor for Virus or Bacteria detection (UVC), and a camera with a CNN tool for item recognition and quantity. Compared different open-source datasets.
Jaipriya et al. [37]	2021	IoT	Designed a framework to ensure that items are trackable and placed correctly on the shelf or else alarm the user by led lamp.
Krishnamoorthy et al. [46]	2021	IoT	Proposed a system can identify the lowest cost of items with the expected delivery time via a loaded list.
Lee et al. [48]	2021	AIoT	Proposed an automatically labeled training data generator method.
Nadar et al. [58]	2021	IoT	Proposed an application with two options: auto-ordering or waiting for user instructions based on the weight of stored items.
Sane et al. [76]	2021	AIoT	Proposed a system with a machine learning classifier (OpenCV in Python) for image recognition along weighing sensor for stored items' quantity.
Sharma et al. [80]	2021	AIoT	Proposed a system that tracks the user and generates an auto-order period based on household behavior.
Nejekar et al. [63]	2022	IoT	Proposed a system with a Blynk app to track stored items and enable viewing images from inside the fridge.

Table 21  
Limitations of and observations from reviewed articles

Article	Limitations/observations
Loh et al. [50]	The proposed system is only conceptual. Transparent items such as water bottles cannot be detected using IR sensors. Item recognition is missing. Reorder point is not automated. Auto alert of system status.
Hong et al. [33]	The proposed system is only conceptual. Not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers. Additional costs such as wastage are not included. Additional assumptions and constraints could be added to make the system more comprehensive and realistic.
Luo et al. [51]	The system is running locally, not fully automated, and applied only for healthcare purposes. Additional input components for system enhancement along with an internet connection to enable remote access.
Konidala et al. [44]	The proposed system is only conceptual. Not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers. Practical implementation of the proposed systems.
Nayak et al. [62]	Transparent items such as water bottles cannot be detected using IR sensors. Additional input components for system enhancement.
Lloret et al. [49]	Additional sensors could be added to help in items' identification.
Rouillard [73]	More utilization of the input information could result in providing system status about inventory level and auto reorder system.
Bostanci et al. [11]	The proposed system could be used in a shared dorm refrigerator as several users are there. For houses, the facial recognition feature might not be as useful. Instead, additional utilization of input information such as inventory level and auto reorder system would be great.
Hou et al. [35]	The proposed system is only conceptual. Also, automation of the tracking system would be great.
Kaldeli et al. [38]	The proposed system is only conceptual. The paper is about smart homes in general. Additional input components with actual implementation would be great.
Sandholm et al. [75]	Status of the inventory level is not mentioned.
Son et al. [82]	The proposed system is only conceptual. Not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Gürüler [30]	The input sensors were not specified.
Kale et al [39]	Automation and additional input sensor to recognize stored items could be more comprehensive.
Osisanwo et al. [65]	The proposed system is only conceptual. Not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Prapulla et al. [69]	Item recognition is missing.
Panchal et al. [67]	Item recognition is missing.
Calegari et al. [12]	The proposed system is only conceptual. Not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Floarea et al. [23]	Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Hachani et al. [31]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Kwon et al. [47]	Additional input sensors could help in tracking the inventory level status.
Esmacili [5]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Edward et al. [19]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with barcode tags. Not all barcode tags are the same.
Goeddel et al. [26]	The proposed system is only conceptual. Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.

Table 21  
(Continued)

Article	Limitations/observations
Hafidh et al. [32]	Additional sensors could be added to help in items' identification.
Qiao et al. [70]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Shama et al. [77]	Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Shweta [81]	Additional input sensors are required as not all items are vegetables and to enable tracking the inventory level.
Wu et al. [89]	The proposed system is only conceptual, and it would be great if it became automated.
Anand et al. [6]	The proposed system is only conceptual, and it would be great if an auto-order system is included.
Fujiwara et al. [24]	Requires specified space for each item.
Hossain et al. [34]	The proposed system is conceptual. An actual implementation could show the efficiency and effectiveness of the proposed system.
Nasir et al. [61]	Proposed system is only applicable for food items that generates gas, other sensors could be used to measure the quantity and be more comprehensive.
Pachón et al. [66]	Training the system with more items and adding additional components for quantity measurement could enhance the proposed system.
Rezwan et al. [71]	Requires specified space for each item. Additional sensors and automation of item recognition could enhance the system.
Zhongmin et al. [93]	The proposed system is only conceptual. Item recognition is missing.
Abdel-Basset et al. [2]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Ahmed and Rajesh [4]	The proposed system is only conceptual. Item recognition is missing.
Barfeh et al. [8]	Requires specified space for each item. Additional sensors and automation of item recognition could enhance the system.
Bayya [9]	Additional input for item recognition is required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Ferrero et al. [22]	The proposed system is only conceptual. Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Gao et al. [25]	Training the system with more items and adding additional components for quantity measurement could enhance the proposed system.
Khan et al. [41]	Training the system with more items and adding additional components for quantity measurement could enhance the proposed system.
Narayan et al. [60]	Requires specified space for each item. Additional sensors and automation of item recognition could enhance the system.
Ringe et al. [72]	The proposed system is only conceptual, and the item recognition method is not mentioned.
Shariff et al. [78]	Additional input sensors are required as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Sharma et al. [79]	Additional input sensors could help in tracking the inventory level status.
Abd Elminam et al. [1]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Avinash et al. [7]	The proposed system is only conceptual. Additional input sensors could help in tracking the inventory level status.

Table 21  
(Continued)

Article	Limitations/observations
Das et al. [16]	Additional input for item recognition is required and another communication medium could make it more comprehensive.
Dong et al. [18]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Kore et al. [45]	The machine learning approach was not specified. Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Mallikarjun et al. [52]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Mohammad et al. [56]	The proposed system is conceptual. Additional input sensors are required to help in tracking the inventory level status could make it more comprehensive as not all items are tagged with RFID tags. Not all RFID tags are compatible with all RFID readers.
Nagarajua et al. [59]	The proposed system is only conceptual. Additional input sensors could help in tracking the inventory level status.
Saha et al. [74]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Velasco et al. [86]	The machine learning approach was not specified. Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Chakilam et al. [13]	Training the system with more items could enhance the proposed system.
Che Soh et al. [14]	The proposed system is conceptual. Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Datey [17]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Gull et al. [27]	Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Gupta et al. [28]	The proposed system is conceptual. Automation and additional input sensor to help in items recognition could make it more comprehensive.
Jain et al. [36]	Training the system with more items could enhance the proposed system.
Jaipriya et al. [37]	Automation and additional input sensor to help in items recognition and tracking the inventory level status could make it more comprehensive.
Krishnamoorthy et al. [46]	Automation and additional input sensor to help in items recognition and tracking the inventory level status could make it more comprehensive.
Lee et al. [48]	The proposed system is conceptual. Automation and additional input sensor to help in tracking the inventory level status could make it more comprehensive.
Nadar et al. [58]	The proposed system is conceptual. Automation and additional input sensor to help in items recognition and tracking the inventory level status could make it more comprehensive.
Sane et al. [76]	The proposed system is conceptual. Automation and additional input sensor to help in items recognition and tracking the inventory level status could make it more comprehensive.
Sharma et al. [80]	The proposed system is conceptual. Automation and additional input sensor to help in items recognition and tracking the inventory level status could make it more comprehensive.
Nejekar et al. [63]	Training the system to be fully automated could enhance the proposed system.

Table 22  
Summarization of used approach and components by the reviewed articles

Reviewed article	Approach		Components																					
	Year (yy)	IoT	AIoT	IR sensors	Ultrasonic sensors	Climate sensor	Light sensor	Gas Sensor	Door Open-Close sensor	Weight sensor	GSM Module	RFID system	Barcode system	Connection medium	Controller <sup>1</sup>	Internet Protocol <sup>2</sup>	Offline-Database <sup>6</sup>	IoT Platform <sup>3</sup>	Touchscreen/PC	Mobile Application	Webcam/Camera module	Recognition module <sup>4</sup>	Implementation (C, S, A) <sup>5</sup>	
Loh et al. [50]	04	X		X					X	X					X									C
Hong et al. [33]	07	X									X				X				X					C
Luo et al. [51]	09	X										X		X		X		X						S
Konidala et al. [44]	11	X									X			X	X	X								C
Nayak et al. [62]	11	X		X					X	X	X			X	X	X			X					S
Lloret et al. [49]	12	X			X	X	X	X	X	X				X		X	X		X	X				A
Rouillard [73]	12	X											X					X						A
Bostanci et al. [11]	13		X												X				X	X	X	X		S
Hou et al. [35]	13	X									X	X	X					X	X	X				C
Kaldeli et al. [38]	13	X							X	X	X								X					C
Sandholm et al. [75]	14		X	X					X							X		X			X	X		A
Son et al. [82]	14	X							X		X								X					C
Gürüler [30]	15	X			X					X				X	X									S
Kale [39]	15	X							X	X	X			X	X									S
Osisanwo et al. [65]	15	X			X						X			X	X	X		X						C
Prapulla et al. [69]	15	X		X					X	X									X	X				S
Panchal et al. [67]	15	X		X	X					X				X	X		X			X				A
Calegari et al. [12]	16	X									X				X					X				C
Floarea et al. [23]	16	X									X				X			X						S
Hachani et al. [31]	16	X									X							X	X	X				C
Kwon et al. [47]	16		X						X					X	X	X		X		X	X	X		S
Esmaili. [5]	17	X							X	X	X				X				X					C
Edward et al. [19]	17	X		X	X							X		X					X					C
Goeddel et al. [26]	17		X							X							X	X			X	X		C
Hafidh et al. [32]	17	X							X	X	X				X									S
Qiao et al. [70]	17	X			X				X		X			X				X		X				C
Shama et al. [77]	17	X		X	X						X			X	X			X	X	X				S
Shweta [81]	17		X												X		X				X	X		S
Wu et al. [89]	17	X		X		X								X	X			X	X	X	X	X		C
Anand et al. [6]	18		X		X			X	X	X	X			X	X			X			X	X		C
Fujiwara et al. [24]	18		X							X				X					X	X		X		S
Hossain et al. [34]	18		X		X			X	X	X	X	X		X	X			X			X	X		C
Nasir et al. [61]	18	X			X			X						X	X		X	X		X				S
Pachón et al. [66]	18		X														X		X		X	X		S
Rezwan et al. [71]	18	X							X					X	X			X		X				A
Zhongmin et al. [93]	18	X			X		X		X					X	X				X					C
Abdel-Basset et al. [2]	19	X									X			X	X	X								C
Ahmed and Rajesh. [4]	19	X			X				X					X	X			X		X				C

Table 22  
(Continued.)

Reviewed article	Year (yy)	Approach		Components																				
		IoT	AIoT	IR sensors	Ultrasonic sensors	Climate sensor	Light sensor	Gas Sensor	Door Open-Close sensor	Weight sensor	GSM Module	RFID system	Barcode system	Connection medium	Controller <sup>1</sup>	Internet Protocol <sup>2</sup>	Offline-Database <sup>6</sup>	IoT Platform <sup>3</sup>	Touchscreen/PC	Mobile Application	Webcam/Camera module	Recognition module <sup>4</sup>	Implementation (C, S, A) <sup>5</sup>	
Barfeh et al. [8]	19	X		X									X	X		X				X			S	
Bayya [9]	19	X		X					X	X	X		X	X									A	
Ferrero et al. [22]	19	X									X						X		X				C	
Gao et al. [25]	19		X																			X	S	
Khan et al. [41]	19		X	X				X	X					X			X			X	X		S	
Narayan et al. [60]	19	X							X				X	X	X				X				A	
Ringe et al. [72]	19	X					X	X					X	X	X		X		X	X			C	
Shariff et al. [78]	19	X		X	X	X	X			X	X			X				X					S	
Sharma et al. [79]	19		X	X				X						X		X				X	X		S	
Abd Elminam et al. [1]	20	X										X	X	X					X				S	
Avinash et al. [7]	20		X										X	X				X	X	X	X		C	
Das et al. [16]	20	X		X	X					X				X		X			X				S	
Dong et al. [18]	20		X			X	X					X		X			X			X	X		S	
Kore et al. [45]	20		X						X				X	X		X	X		X	X	X		S	
Mallikarjun et al. [52]	20		X	X	X				X				X	X	X		X		X	X	X		A	
Mohammad et al. [56]	20		X					X			X		X	X						X	X		C	
Nagarajua et al. [59]	20		X											X			X			X	X		C	
Saha et al. [74]	20		X											X					X	X	X		S	
Velasco et al. [86]	20	X			X			X	X				X	X			X		X	X			S	
Chakilam et al. [13]	21		X			X	X	X					X	X			X	X		X	X		S	
Che Soh et al. [14]	21		X														X			X	X		C	
Datey [17]	21		X					X	X	X			X	X	X		X		X	X	X		S	
Gull et al. [27]	21		X					X	X				X	X				X				X	S	
Gupta et al. [28]	21	X				X			X					X					X				C	
Jain et al. [36]	21		X			X	X	X						X			X		X	X	X		S	
Jaipriya et al. [37]	21	X			X					X	X			X				X					S	
Krishnamoorthy et al. [46]	21	X			X	X		X		X			X	X	X		X		X	X			S	
Lee et al. [48]	21		X					X												X	X		C	
Nadar et al. [58]	21	X				X			X				X	X			X		X				C	
Sane et al. [76]	21		X			X			X				X	X			X	X	X	X	X		C	
Sharma et al. [80]	21		X					X						X		X			X	X	X		C	
Nejekar et al. [63]	22	X		X	X	X	X	X	X	X	X			X					X	X			S	
Unified Framework	23		X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	X	C

<sup>1</sup>Several types of controllers used in reviewed papers: Raspberry Pi, Arduino (Uno, WeMos D1 R2, and ATmega 2560), etc.<sup>2</sup>Two internet protocols used in reviewed papers: IP address and MQTT.<sup>3</sup>IoT platforms refer to different types used in reviewed papers: Cloud-based, Google Firebase, Thingspeak, etc.<sup>4</sup>Recognition module referred to machine learning or deep learning to recognize images and data used in reviewed papers.<sup>5</sup>Implementation: C for conceptual, S for simulation, and A for the actual implementation of the proposed system by the reviewed papers.<sup>6</sup>Different types of databases were used in the reviewed papers: PLX-DAQ, Excel spreadsheet, etc.

Table 23  
Frequency of used components in the introduced systems by the reviewed papers

	IoT	AIoT	IR sensor	Ultrasonic	Climate sensor	Light sensor	Gas sensor	Door sensor	Weight sensors	GSM Module	RFID Sys.	Barcode Sys.	Connection Medium	Controller	Internet protocol	Offline-Database	IoT Platform	Tablet/PC	Mobile Application	Webcam	Camera module	Recognition module	Conceptual	Simulation	Actual
IoT	43	0	12	7	14	2	6	10	15	14	18	5	23	34	7	7	15	16	24	4	0	0	20	17	6
AIoT	0	27	4	3	4	2	4	9	11	1	3	2	10	19	4	7	14	8	11	1	23	27	10	15	2
IR sensor	12	4	16	4	4	1	2	5	5	6	4	1	6	14	3	4	5	5	9	3	3	4	3	9	4
Ultrasonic	7	3	4	10	3	1	3	4	5	5	4	1	5	9	3	4	4	4	6	1	3	3	2	5	3
Climate sensor	14	4	4	3	18	3	7	2	9	3	5	2	12	16	2	3	10	7	12	1	4	4	8	9	1
Light sensor	2	2	1	1	3	4	2	1	1	0	0	1	3	3	1	1	3	3	2	1	2	2	1	2	1
Gas Sensor	6	4	2	3	7	2	10	1	7	1	3	0	7	9	2	2	5	5	5	2	3	4	3	6	1
Door sensor	10	9	5	4	2	1	1	19	7	8	5	1	8	14	5	5	5	7	6	2	8	9	8	9	2
Weight sensors	15	11	5	5	9	1	7	7	26	4	5	1	17	21	5	4	14	7	17	3	9	11	10	11	5
GSM Module	14	1	6	5	3	0	1	8	4	15	5	1	7	13	1	4	1	7	4	0	1	1	3	9	3
RFID System	18	3	4	4	5	0	3	5	5	5	21	2	8	15	4	0	9	9	7	1	3	3	15	5	1
Barcode sys.	5	2	1	1	2	1	0	1	1	1	2	7	2	5	1	1	4	2	3	0	2	2	3	3	1
Connection Medium	23	10	6	5	12	3	7	8	17	7	8	2	33	29	7	7	16	13	22	3	8	10	13	15	5
Controller	34	19	14	9	16	3	9	14	21	13	15	5	29	53	9	11	20	16	28	4	18	19	20	28	5
Internet protocol	7	4	3	3	2	1	2	5	5	1	4	1	7	9	11	1	6	2	5	2	3	4	5	2	4
Offline-Database	7	7	4	4	3	1	2	5	4	4	0	1	7	11	1	14	3	5	9	0	7	7	2	10	2
IoT Platform	15	14	5	4	10	3	5	5	14	1	9	4	16	20	6	3	29	6	17	4	13	14	15	10	4
Tablet/PC	16	8	5	4	7	3	5	7	7	7	9	2	13	16	2	5	6	24	12	1	6	8	10	13	1
Mobile Application	24	11	9	6	12	2	5	6	17	4	7	3	22	28	5	9	17	12	35	4	10	11	14	16	5
Webcam	4	1	3	1	1	1	2	2	3	0	1	0	3	4	2	0	4	1	4	5	0	1	2	2	1
Camera module	0	23	3	3	4	2	3	8	9	1	3	2	8	18	3	7	13	6	10	0	23	23	10	12	1
Recognition module	0	27	4	3	4	2	4	9	11	1	3	2	10	19	4	7	14	8	11	1	23	27	10	15	2
Conceptual	20	10	3	2	8	1	3	8	10	3	15	3	13	20	5	2	15	10	14	2	10	10	30	0	0
Simulation	17	15	9	5	9	2	6	9	11	9	5	3	15	28	2	10	10	13	16	2	12	15	0	32	0
Actual	6	2	4	3	1	1	1	2	5	3	1	1	5	5	4	2	4	1	5	1	1	2	0	0	8

using the AIoT approach. The GSM module communicates with the grocery to authenticate orders [16]. The systems applying IoT technology tend to be more robust than in the past. Based on the reviewed papers, the reason is that AIoT requires additional components that need more time for training data, effort, and cost as it involves many steps. Looking at the implementation of the introduced systems will clear that reason, as shown in Fig. 4. The AIoT approach in the reviewed papers used a prototype by a computer to test the results and prove the system's accuracy as it is all about computational tools rather than conceptual or actual implementations. In contrast, IoT is based on the components' functionalities when connected locally and to the internet. Therefore, it is possible with the IoT approach to predict the results conceptually rather than through simulation or actual implementation.

Table 23 shows the frequency of approaches and components used in the introduced systems by the reviewed papers. The dark diagonal line indicates the frequency of single elements. The rest of the numbers indicate the pairwise frequency between the used components and approaches in the reviewed papers. The papers proposed their systems were 88.57% conceptual and simulated systems, while only 11.43% were actually implemented systems. It indicates that the implementation of such systems to gather practical results might be expensive. Especially when it comes to the implementation of AIoT systems, which involves machine learning that requires a lot of time and money.

Table 24  
Comparison between the used components by the reviewed papers

Component	Functionality and advantages	Limitations and disadvantages
IR sensor	<ul style="list-style-type: none"> <li>- Senses liquid levels and measures distance.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- Not accurate as non-transparency items and can block it.</li> <li>- Transparency items like water bottles cannot be measured.</li> <li>- Does not identify objects.</li> </ul>
Ultrasonic sensor	<ul style="list-style-type: none"> <li>- Measures the distance between objects.</li> <li>- Transparency items like water bottles cannot be measured.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- Does not identify objects.</li> </ul>
Climate sensors	<ul style="list-style-type: none"> <li>- Measures the temperature and humidity, and some types allow control from a distance.</li> <li>- Important to ensure food freshness.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- N/A</li> </ul>
Light sensors	<ul style="list-style-type: none"> <li>- Senses the light to trigger other components.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- N/A</li> </ul>
Gas Sensor	<ul style="list-style-type: none"> <li>- Senses and measures the gas generated from organic stored items.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- Does not identify objects.</li> <li>- Only for organic items.</li> </ul>
Door Open-Close sensors	<ul style="list-style-type: none"> <li>- Senses compartment door movement and gives alerts accordingly.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- N/A</li> </ul>
Weight sensors	<ul style="list-style-type: none"> <li>- Weigh the items placed on it.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- Does not identify objects.</li> <li>- Each stored item has to have a sensor.</li> </ul>
Mobile Application GSM Module	<ul style="list-style-type: none"> <li>- Allows control, monitor, and many other features from a distance.</li> <li>- Enables communication with the smart refrigerators' connected devices via a telecommunication medium.</li> <li>- Cheap, compatible with most controllers.</li> </ul>	<ul style="list-style-type: none"> <li>- Might have a compatibility issue between iOS and Android.</li> <li>- Required SIM card.</li> <li>- Cellular coverage vary.</li> <li>- Price may also vary.</li> </ul>
RFID System	<ul style="list-style-type: none"> <li>- Some models scan items automatically.</li> <li>- Items identification.</li> <li>- Low maintenance required.</li> <li>- Some tags do not require a battery and are cheap.</li> <li>- Tracks items by recognizing added items and Measures items' remaining quantities.</li> </ul>	<ul style="list-style-type: none"> <li>- Some models required manual scanning.</li> <li>- Some materials can block it.</li> <li>- Does not measure levels of liquid items.</li> <li>- Does not identify un-tagged items.</li> <li>- Many models exist, so it does not identify items with unsupported tags.</li> <li>- It is subjected to items' information stored in the tags.</li> </ul>

Table 24  
(Continued.)

Component	Functionality and advantages	Limitations and disadvantages
Barcode system	<ul style="list-style-type: none"> <li>- Items identification.</li> <li>- Tracks added items.</li> </ul>	<ul style="list-style-type: none"> <li>- Requires a manual scanning of items before entering the storage compartment.</li> <li>- Requires additional components and methods to track taken items.</li> <li>- Information of items in the barcode varies from type to type.</li> <li>- Not all items are tagged with a barcode.</li> </ul>
Connection medium	<ul style="list-style-type: none"> <li>- Enables communication channel.</li> <li>- Easy to configure and established [22].</li> </ul>	<ul style="list-style-type: none"> <li>- Requires special component.</li> <li>- There might be security issues.</li> </ul>
Controller	<ul style="list-style-type: none"> <li>- There are many types of controllers used in the reviewed papers, so the functionalities vary between them.</li> </ul>	<ul style="list-style-type: none"> <li>- It depends on the chosen model.</li> </ul>
Internet protocol	<ul style="list-style-type: none"> <li>- Method to exchange data and information between devices and the internet.</li> </ul>	<ul style="list-style-type: none"> <li>- Some compatibilities issues might occur.</li> <li>- Some protocols might be intercepted and causes security issues.</li> </ul>
Offline-Database	<ul style="list-style-type: none"> <li>- Stores text or image data or both.</li> <li>- Secure if not connected to a network and exposed to the internet.</li> </ul>	<ul style="list-style-type: none"> <li>- No remote access.</li> </ul>
IoT Platform	<ul style="list-style-type: none"> <li>- Some platforms are capable of data processing.</li> <li>- May replace many components like the microcontroller, database, PC applications.</li> </ul>	<ul style="list-style-type: none"> <li>- It requires an internet connection.</li> <li>- There might be security issues.</li> </ul>
Tablet/Touchscreen/PC	<ul style="list-style-type: none"> <li>- Allows monitoring, controlling, and many other features.</li> <li>- Some have built-in voice capturing.</li> </ul>	<ul style="list-style-type: none"> <li>- The attached ones are not mobile.</li> <li>- Adding more expenses.</li> </ul>
Webcam/Camera module	<ul style="list-style-type: none"> <li>- Enables capturing images.</li> <li>- Some cameras have high resolutions and 360-degree feature.</li> <li>- Better to use for image recognition.</li> </ul>	<ul style="list-style-type: none"> <li>- Compatibility issues.</li> <li>- Images might not be apparent if one item blocks another; adding more cameras adds expense.</li> <li>- Might face compatibility issues.</li> <li>- Requires time, effort, well-trained database, processor, and programming skills for image recognition.</li> <li>- Does not identify undefined items.</li> </ul>
Recognition module	<ul style="list-style-type: none"> <li>- Enables image processing, recognition, and facial recognition.</li> <li>- Reduces human interaction with the devices.</li> <li>- Helps in the decision-making process.</li> </ul>	<ul style="list-style-type: none"> <li>- Requires camera module.</li> <li>- Requires time, effort, well-trained database, processor, algorithms, and programming skills for image recognition.</li> <li>- Does not identify undefined items.</li> </ul>

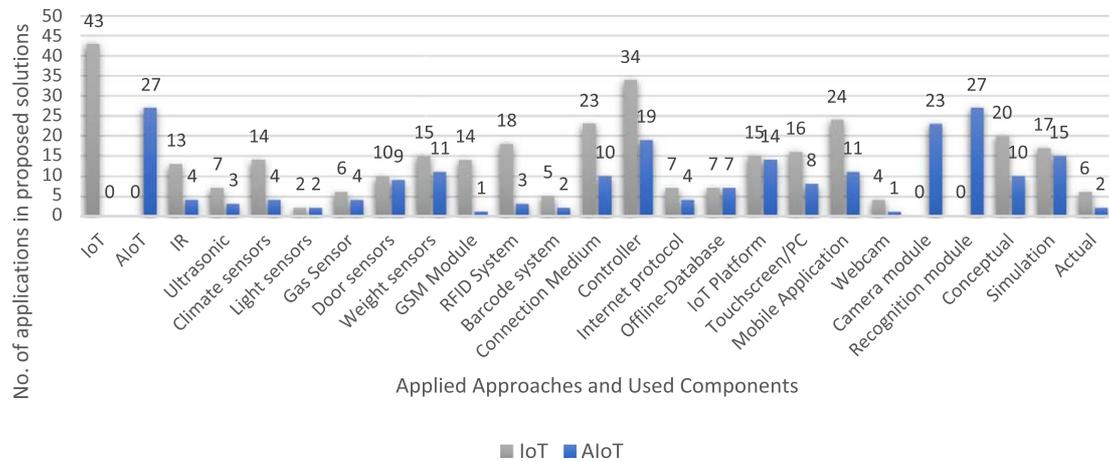


Fig. 4. IoT vs. AIoT approach on smart storage compartments.

#### 4. Findings and discussion

After analyzing the systems introduced in the 70 reviewed papers, this section will discuss the findings based on the descriptive analysis. Therefore, this section will have two sub-sections. The first section is about the used components and their functionalities, advantages, limitations, and disadvantages regarding the identification of stored items, as shown in Table 24. Finally, based on Table 22 and correlation analysis results Table 23, the decision tree will help develop a unified framework in which the used components are connected and communicate with each other to enhance the household management system performance.

##### 4.1. Components comparison

In this section, a comparison between the used components in the introduced systems in the reviewed papers is discussed. Table 24 highlights the functionality and advantages of each component and the limitations and disadvantages as stated in the reviewed papers. Some components have nothing to do with tracking food items. Therefore, N/A will be written as its limitations and disadvantages.

##### 4.2. A unified framework

Based on the previous analysis of the reviewed literature, this paper presents a unified framework in which all possible combinations of the mentioned components for the household replenishment system are presented. A decision-tree technique that uses a continuous decision variable, and is known as a regression tree, is used to connect and allow the components to communicate [54]. The unified framework is a showcase that gives researchers and manufacturers a starting point for enhancing the performance of the household replenishment system shown in Fig. 6. The components are categorized based on their functionalities, as shown in Table 25 followed by Table 26 for a description of how they work.

Therefore, the researchers and manufacturers would gain knowledge over the years about the components used in household replenishment systems based on the reviewed papers. For example, Loh et al. proposed a design of a system that can keep track of the free space inside the storage compartment. The authors divided the design into stages. Starting from the sensing sensors, through control and interface circuits, and ending with GSM circuit. The designed system was dedicated for space measurement and tracking food quantity [50]. Nayak et al. [62] and Panchal et al. [67] proposed a framework that presents specific components for remote monitoring. The framework was presented as a block diagram starting from the sensors ending with the market. The system was dedicated to monitor the stored items via IR sensor and initiate orders to the nearest market [62,67]. Hou et al. [35] proposed a framework that contains barcode and RFID technologies for food management. The proposed framework was split

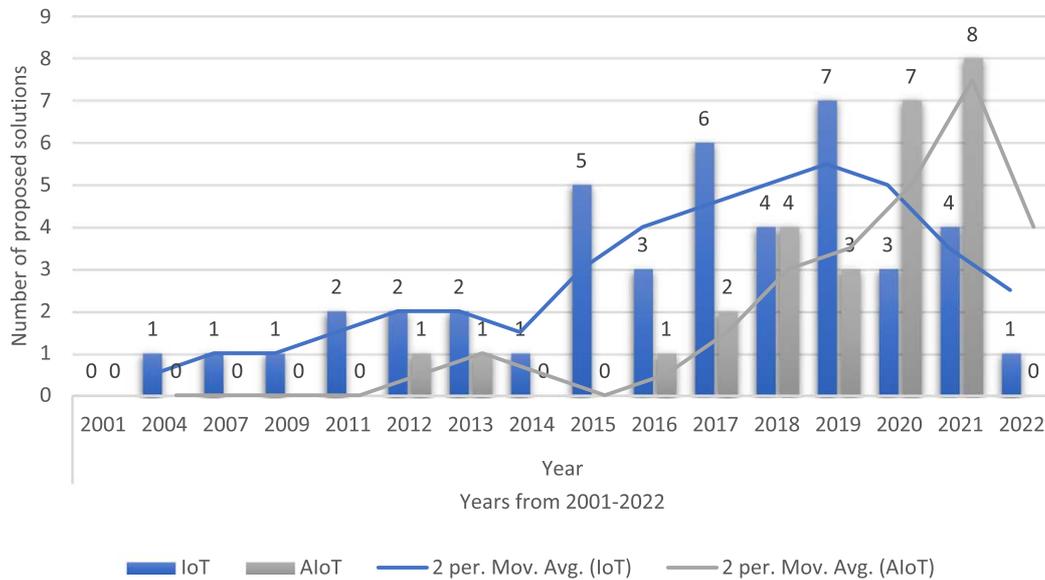


Fig. 5. Application of IoT and AIoT over the years.

into several stages called units. The sensing, storage, control, push, and display units. The system was dedicated to the tagged items, and the non-tagged items can be entered manually [35]. Gürüler [30] proposed a block diagram that also shows specific components of the system. The proposed system was designed to allow the household to communicate with the storage compartment via SMS [30]. Hachani et al. [31], Floarea et al. [23], and Esmaeili [5] proposed a system architecture that uses RFID technology to capture stored items' information. Kwon et al. proposed a system architecture that contains three stages. The first stage consists of capturing sensors. The second stage consists of database management. The final stage is the output stage [47]. Edward et al. [19] proposed a system architecture that consists of input, processing, connection medium, and output [19]. Wu et al. [89] proposed a system architecture that shows the identification of the items via Google Firebase using inside and outside cameras [89]. Anand et al. [6] have a slight difference. Anand et al. [6] added more sensors like gas, ultrasonic, and weighing sensors connected to a controller and to Google Firebase [6]. Nasir et al. [61] proposed a framework that shows three stages of the system: input, processing, and output. The proposed system was limited to specific components [61].

The unified framework presented in this paper in Fig. 6 is meant to be a generalized form of a framework derived from the specified frameworks presented in the reviewed papers. It is divided into four stages, and each stage consists of a combination of all components used in the reviewed papers. The starting triggers, input, processing, and output stages. The starting triggers stage consists of either automated or manual components that trigger the input stage components to start to work. The input stage's components capture the data from the storage compartment and pass it through to the processing stage. The processing stage then processes the captured data into meaningful information and displays them accordingly via the output stage's components. Later, the researchers and manufacturers could modify the components based on their perspectives and objectives. Figure 6 assigns a number to each component to help with components' illustrations in Table 26.

However, the systems with the AIoT approach were developed in 2012 and 2013, then stopped until 2016, when they resumed growing. In 2020 and 2021, the use of AIoT approaches doubled the uses of IoT, Fig. 5, because of the technological revolution in Artificial Intelligence and human-less interactions with machines.

Based on literature, Table 25 classifies each component shown in Fig. 6 with its category. The categories are classified based on the components functionalities, as follows: 1) starting triggers are the components that trigger all input components to work except the barcode scanner and some RFID scanners that could work manually; 2) input components can be used for both quality management and quantity management, the quality management components are used to monitor the environment inside the storage compartment, the quantity management components

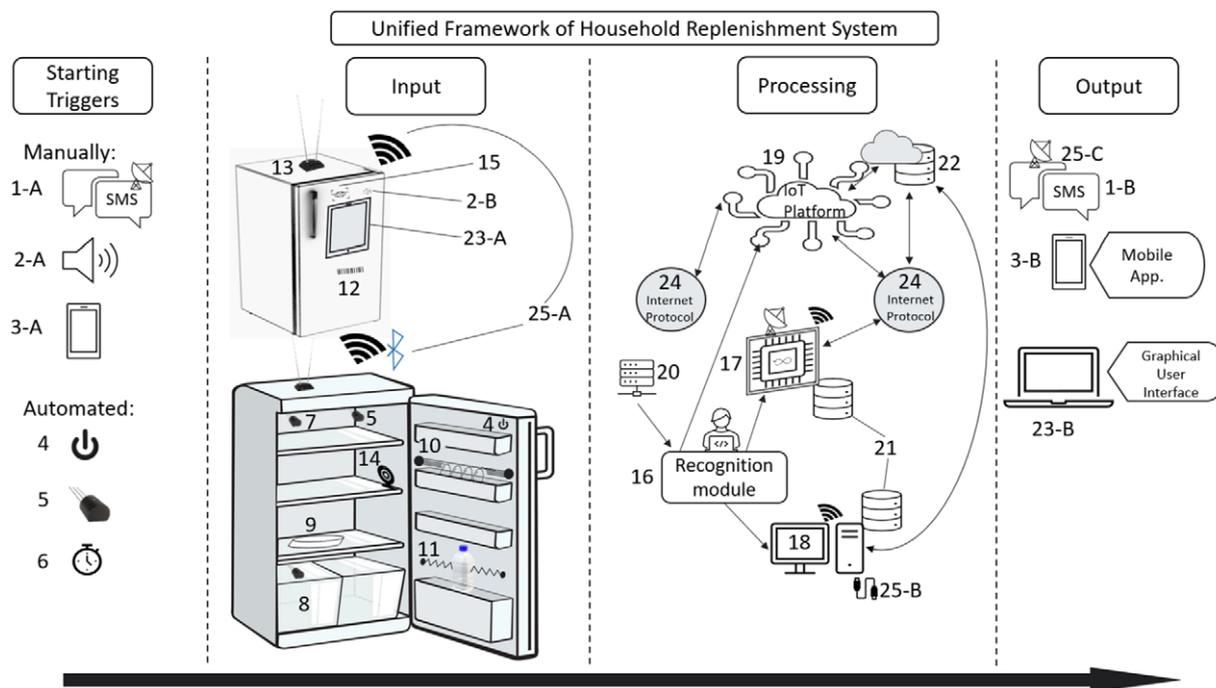


Fig. 6. A unified framework for household replenishment system.

Table 25  
Categories of used components by the reviewed papers

Starting triggers	Input	Processing	Output	Storage and connectivity
User Commands	Gas Sensor	IoT Platform	Mobile App.	Offline database
Light Sensor	Climate Sensor	Controllers	Laptop	Online database
Door Sensor	IR Sensors	Desktop computer	Attached Tablet	Bluetooth
Timer	Ultrasonic	Attached Tablet	Desktop computer	Ethernet
	Weighing Sensor	Recognition Module		WiFi
	RFID System	eNose System		Internet Protocol
	Barcode Sensor			GSM Module
	Camera Module			

are used to keep track of the stored items’ quantities and the household consumption habit, some of the quantity management components can identify the stored items and some do not; 3) the processing components receive the data from the input components to be processed and recorded; 4) then, the processed information can be retrieved by the output components to be displayed; 5) the processed information can be stored in either online or offline databases, and the connection medium that enables communication between all system components could vary.

Based on literature, a detailed illustration of the unified framework components shown in Fig. 6 and tied to their categories and associated with the assigned numbers is provided in Table 26.

### 5. Conclusion and future research directions

This paper focused on smart refrigerators as they always contain items that require monitoring. In contrast, other home appliances, such as washers and dryers, are used occasionally. However, other things could be learned from other smart appliances at home—for example, washer, dryer, coffee machine, etc. Therefore, in the future, there

Table 26  
Detailed illustration of the framework components

Category	Component	Illustration
Starting Triggers Manually	(1-A) User Commands	The system receives user commands via GSM module (SMS) [50]
	(2-A) Voice Commands	The system can be woken up via household voice either from a smartphone or the built-in voice feature in the attached tablet (23-A) [31].
	(3-A) Pushbullet	A Pushbullet in the mobile application for the household to trigger the system to seek updates [61].
Starting Triggers Automated	4- Door open-close sensor	Once the storage compartment door is opened, the sensor triggers the input components to work [62].
	5- Light sensor	Senses when the light is turned on inside the storage compartment. It triggers the input components to work [18].
	6- Timer	A timer can be programmed to trigger the input components to work [9].
Input components Quality Management	7- Climate sensor	Monitors the humidity and the temperature inside the refrigerator. It could have a feature that controls the temperature based on household preferences [76].
	8- Gas sensor	Senses gas generated by organic food items such as fruit, vegetables, meat, etc., then updates the system [36].
Input components Quantity Management With out identification	9- Weight sensor	Measures and keeps track of the weight of stored item [36].
	10- IR sensor	Measures and keeps track of the distance between stored items and the untransparent liquid level [89].
	11- Ultrasonic sensor	Measures and keeps track of the distance between stored items and the transparent liquid level [6].
Input components Quantity Management With identification	12- Barcode reader	The household manually scans barcodes of the stored items while placing them in the storage compartment. (2-B) a built-in microphone in the attached tablet or (3-B) the microphone in the smartphone could be used to manually identify items by speaking their names while placing them in the storage compartment [24,34].
	13- RFID scanner	Updates and monitors the RFID-tagged items inside the storage compartment [44].
	14- Inside camera	Captures images of the stored items and sends them to the controllers to be identified and processed. In some cases, the images go directly to the household to monitor from a distance [11]. The inside camera is also used to track which food items have been taken by the household from inside the storage compartment to track the user consumption habit [80].
	15- Outside camera	The outside camera in the attached tablet (23-A) is used for facial recognition to recognize the household member opening the refrigerator and track the consumption habit [80].

Table 26  
(Continued.)

Category	Component	Illustration
Processing components Controllers and Methods	16- Recognition module	Recognition module is a processing method that is an algorithm using artificial intelligence techniques. It processes images captured by (14) the inside camera, (15) the outside camera, natural language from (2-B) and (3-B) microphones, and the data collected by (20) the eNose system. The module can be located in (17) the controller, (18) a home desktop computer, and (19) an IoT platform [11,27,75,80].
	17- Controller	A controller is a device (e.g., Arduino) that receives data from all (input components), stores them in (21) the offline database or (22) the online database, and processes them. Then, it sends the processed data to the household. The household could reconfigure the controller via (18) the home desktop computer, (3-B) the smartphone mobile application, or (23-B) the GUI in the laptop [44,51,67].
	18- Desktop computer	The home desktop computer acts as an offline processor to collect data from all input components, store them in (21) the offline database, and process them. Then, it sends the processed data to the household. (23-A) The attached tablet works similarly. The household could edit the processed data directly [33].
	19- IoT platform	The IoT platform has several uses. One, it acts as (17) the controller and (18) the home desktop computer, but in a larger capacity and completely online. It uses (22) an online database to store and process the data. It enables the household to access information from a distance using (3-B) the smartphone mobile application or (23-B) the GUI in the laptop [13].
	20- eNose system	The eNose system collects the data from (8) gas sensors to send them to the controllers with an embedded (16) recognition module [27].
Output components	18- Desktop computer	The home desktop computer with a graphical user interface is also used locally as an output component that enables the household to view information about the system [27].
	(23-A) Attached tablet	The attached tablet is locally used via an installed mobile application to view information about the system [46].
	(3-B) Smartphone	Smartphone could be used via an installed mobile application to remotely view the system's information or receive SMS messages via the GSM module [17].
	(23-B) Laptop	Laptop could be used as mobile device with a graphical user interface to remotely view information about the system [89].
Storage and Connectivity components	21- Local database	An offline database that could be used to store the processed data by (18) the home desktop computer or (17) the controller locally. It enables the household to manage it physically using (18) the home desktop computer [61].
	22- Online database	The processed data could be stored Online by (18) the home desktop computer or (17) the controller over the internet. The household can access the data locally through (18) the home desktop computer or from a distance using (3-B) the smartphone mobile application or (23-B) the GUI in the laptop [13].
	24- Internet	The Internet Protocols govern the connection between the controllers and the internet [34].
	(25-A) WiFi	Enables some components to communicate and connect to the internet [77].
	(25-B) Bluetooth	Enables some components to communicate [1].
	(25-C) Ethernet	Enables some components like (18) the home desktop computer to the internet [30].

might be a need to conduct a review of other home appliances and show their effect under the smart home umbrella in general. This paper surveyed the relative work conducted on IoT and AI technologies applications on a home perishable items storage compartment to convert them into smart ones and show the households' interaction. It focused on papers starting from 2000 using the mentioned keywords and criteria in methodology Section 2. A total of 70 papers were selected to be reviewed that applied either IoT or AIoT technologies toward smart refrigerators and/or cabinets following PRISMA search strategy. Twenty-seven papers shed more light on the combination of AIoT components with storage compartments for perishable items. Ten papers showed the results based on conceptual assumptions, fifteen papers based on simulations, and two papers on the implementation of the system. Forty-three papers drew the connection between IoT components and storage compartments for perishable items. Twenty papers showed the solutions with conceptually assumed results, seventeen of them were based on simulation, and six were on implementation of the system. These results indicate that the verification and validation of AIoT applications tend to be expensive. Therefore, this paper uses data mining techniques to build a continuous decision variable to analyze the reviewed papers that resulted in a unified framework including all possible combinations of used components. The used components were categorized into stages based on their functionalities. The starting triggers, input, processing, and output stages. A detailed description of the components used, functionalities, and limitations was provided.

The systematic review starts with section one, the introduction; section two, the research methodology; section three, the Household Replenishment Systems Analysis; section four, the findings and discussions; and ends with the conclusion and future directions section.

This comprehensive research pointed out the reviewed papers' approaches, contributions, used components, and limitations.

The proposed framework was presented in this systematic review. It could be considered a showcase that gives researchers and manufacturers a starting point for enhancing the performance of the household replenishment system. It provides a visualization of all possible combinations of components used in literature for such a matter. Verification and validation to demonstrate the efficiency and effectiveness of the unified framework need to be addressed in the future.

Several summarized tables of the conducted review were provided. The summarized tables are intended to give researchers and manufacturers many factors to decide on the most effective combination of components that could be included or excluded with respect to their objectives. Moreover, the summarized tables could be used as guidelines to assign weight coefficients among the presented components based on their trendiness, frequency, functionalities, etc.

To the best of the authors' knowledge, this would be the first comprehensive approach that includes all possible combinations of components used in household replenishment systems. Based on the thoroughness of the reviewed literature, the authors believe that this is a comprehensive framework expected to yield better results, but that is something that needs to be explored in other opportunities. In the spirit of continuous improvement, testing will result in outcomes that might lead us to make changes. More investigation needs to be done to include dissertations, theses, and books that tackle the same topic. Future work is to enlarge the scope and include introduced systems to enhance the household replenishment system to reduce food wastage in dissertations, theses, books, and patents. Furthermore, further investigation into smart systems and smart home appliances would be an introduction to extensive approaches like smart homes, markets, healthcare divisions, industries, and eventually smart cities.

Finally, this paper provides future research directions and sheds more light on areas of improvement for manufacturing companies gathered from the reviewed papers to enhance the household replenishment system. The proper use of IoT and AI technology may improve the household replenishment system in many ways:

- Improvement of household replenishment system.
- Ways to find an optimal threshold period.
- The auto setting of the threshold period by the system.
- Prediction of patterns of user habits based on consumption and purchases.
- Criteria of store selection for the replenished items.
- Remote maintenance for smart refrigerator.
- Considering using components by maintaining the minimal cost of implementation.

- Investigate and resolve security issues.
- Expand the search by involving vendor management inventory (VMI).
- Consider the involvement of the suppliers to enhance the online shopping system to be automatically done.
- Expand the search to identify the significant IoT and AIoT components that work together for substantial results to enhance the household replenishment system and reduce food wastage along with verification and validation.

## Conflict of interest

None to report.

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