

Wireless Energy Harvesting for the Internet of Things

Pouya Kamalinejad, Chinmaya Mahapatra, Zhengguo Sheng, Shahriar Mirabbasi, Victor C. M. Leung, and Yong Liang Guan

ABSTRACT

The Internet of Things (IoT) is an emerging computing concept that describes a structure in which everyday physical objects, each provided with unique identifiers, are connected to the Internet without requiring human interaction. Long-term and self-sustainable operation are key components for realization of such a complex network, and entail energy-aware devices that are potentially capable of harvesting their required energy from ambient sources. Among different energy harvesting methods, such as vibration, light, and thermal energy extraction, wireless energy harvesting (WEH) has proven to be one of the most promising solutions by virtue of its simplicity, ease of implementation, and availability. In this article, we present an overview of enabling technologies for efficient WEH, analyze the lifetime of WEH-enabled IoT devices, and briefly study the future trends in the design of efficient WEH systems and research challenges that lie ahead.

INTRODUCTION

The Internet of Things (IoT) is an intelligent infrastructure of uniquely identifiable devices capable of wirelessly communicating with each other, services, and people on a large scale through the Internet [1]. IoT aims to make the Internet ubiquitous and pervasive, and has the potential to affect many aspects of users' quality of life. The networked heterogeneous devices connected in an IoT structure are typically equipped with sensors, controlling processors, wireless transceivers, and an energy source (e.g., a battery) to monitor their environment and send/receive data. Applications envisioned for IoT span a wide range of fields including home automation, healthcare, surveillance, transportation, smart environments, and many more. One of the dominant barriers to implementing such a grandiose scheme is supplying adequate energy to operate the network in a self-sufficient manner without compromising quality of service (QoS). Therefore, it is imperative to improve the energy efficiency and longevity of devices in IoT.

Although there are numerous methods to achieve energy efficiency, such as using lightweight communication protocols [2] or adopting low-power radio transceivers [3], the recent technology trend in energy harvesting provides a fundamental method to prolong battery longevity. Thus, energy harvesting is a promising approach for the emerging IoT [4]. Practically, energy can be harvested from environmental sources: thermal, solar, vibration, and wireless RF energy sources [5]. While harvesting from the aforementioned environmental sources is dependent on the presence of the corresponding energy source, RF energy harvesting provides key benefits in terms of being wireless, readily available in the form of transmitted energy (TV/radio broadcasters, mobile base stations and handheld radios), low cost, and small form factor implementation. This article presents an overview of wireless energy harvesting units in the context of wireless energy harvesting IoT (WEH-IoT) systems. In this scenario, multiple sensor nodes typically transmit data to a common sink node. The sink node, also known as a gateway, is connected to the network and is accessible to the outside world over the Internet.

A WEH-enabled sensor device usually consists of an antenna, a transceiver, a WEH unit, a power management unit (PMU), a sensor/processor unit, and possibly an onboard battery. Among those components, there are two essential units for energy harvesting: unit and PMU:

- The WEH unit is in charge of harvesting the RF energy and producing a stable energy source for the rest of the device. It also interfaces with the PMU.
- The PMU controls the transceiver, sensing unit functionality, and manages the energy consumption of each unit and/or accommodates battery charging using the harvested energy.

In this article, we also focus on the enabling technologies including high-efficiency wireless/RF energy harvesting rectifiers and low-power wake-up radio for WEH units. We propose a PMU architecture that accommodates a battery charging scheme using the harvested energy through a WEH unit. Furthermore, we analyze

Pouya Kamalinejad, Chinmaya Mahapatra, Shahriar Mirabbasi, and Victor C. M. Leung are with the University of British Columbia.

Zhengguo Sheng is with the University of Sussex.

Yong Liang Guan is with Nanyang Technological University.

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the lifetime of the proposed WEH-IoT system in the context of two common scenarios in IoT networked systems. The energy cost model is described using uniform and random distribution topology of sensor devices. It is shown that the lifetime is increased substantially using the wireless harvesting techniques. Finally, we conclude with a discussion of future research challenges.

THE WIRELESS ENERGY HARVESTING UNIT

The WEH receives the transmitted radio waves with an antenna and converts the received RF energy into a stable direct current (DC) energy source to supply the sensor device. Generally, in the context of IoT, wireless sensor networks (WSNs), and RF identification (RFID) tags, wireless energy sources can be classified into two categories [6]:

1. *Dedicated source*: Dedicated RF sources are deployed to provide a predictable energy supply to the device. Dedicated sources can be optimized in terms of frequency and maximum power to meet the requirements of the sensor devices. A sink node is an example of a dedicated source.
2. *Ambient source*: This type of source is further divided into two subcategories:
 - *Static or anticipated ambient sources*, which are transmitters that radiate stable power over time, are not optimized (e.g., in terms of frequency and transmitted power) to supply the sensor device. Mobile base stations, and broadcast radio and TV are examples of anticipated ambient sources.
 - *Dynamic or unknown ambient sources*, which are transmitters that transmit periodically in a fashion not controlled by the IoT system. Harvesting energy from such sources require an intelligent WEH to monitor the channel for harvesting opportunities. WiFi access points, microwave radio links and police radios are examples of unknown ambient sources.

Different ambient sources transmit at different frequency bands. Harvesting wireless energy at multiple frequency bands complicates the antenna geometry requirements and demands a sophisticated power converter. Therefore, WEH is typically optimized to harvest from the dedicated energy source (e.g., sink node) and may be devised so as to allow ambient energy harvesting as an auxiliary source.

RF-TO-DC RECTIFIER

In practice, the conversion from the received RF power to the usable DC supply comes with a certain amount of power loss in the matching circuit and in the internal circuitry of the power converter. The power conversion efficiency (PCE) of the converter is the ratio of the generated usable DC output power to the input RF power. State-of-the-art RF-to-DC converters (also known as rectifiers) can achieve high PCE values, up to 70 percent or more [7]. PCE is an indication of the amount of harvested energy that is available for the sensor device. The avail-

able harvested power, P_H , is given by a Friis equation [7] and is directly proportional to the transmitted power, P_T , path loss, P_L , transmitter antenna gain, G_T , receiver antenna gain, G_R , power conversion efficiency of the converter, PCE_H , and the square of the wavelength, λ , and is inversely proportional to the square of the communication distance, r , between the source and the device (Fig. 1).

A schematic diagram of the WEH unit is shown in Fig. 1a. The transmission power, communication medium, antenna gains, and frequency of operation are typically dictated by the application requirements. Therefore, a viable design parameter to enhance the harvested power P_H or maximize the communication distance r is PCE. The PCE curve as a function of distance/input power level for a typical rectifier is shown in Fig. 1b. As shown, the PCE is optimized to peak at a certain input level that corresponds to a specific distance (i.e., PCE_{MAX} at r_{opt}).

In addition to a high PCE, other important characteristics of the WEH unit include high sensitivity, wide high-efficiency range, multi-band operation, and ease of implementation. Extensive studies have been performed on techniques to improve the efficiency of the converter unit [7]. One of our recent efforts to enhance PCE for rectifiers operating at small input levels and a technique to enable harvesting RF energy at multiple frequencies with a single antenna (to facilitate energy harvesting from ambient sources) are described in [8].

WAKE-UP RADIO SCHEME

The radio transceiver is typically the most power hungry block of a wireless sensor device. Although the transceiver is rarely called into action during each operation cycle, it has to keep monitoring the channel. This idle listening process is a significant contributor to the overall power consumption of the sensor device.

An efficient approach to address the idle-mode energy consumption is *duty-cycling*, in which the receiver on-demand switches between listening and sleeping states. Among the different categories of duty-cycling (synchronous, pseudo-asynchronous, and pure asynchronous), the latter provides the most efficient solution in terms of energy consumption [3].

In the asynchronous approach, the sensor device is in deep sleep mode and only wakes up when signaled by the sink node or its neighboring devices through an interrupt command generated by a low-power wake-up radio (WUR). The timing diagram of the asynchronous communication approach is shown in Fig. 1c. Since the WUR is constantly active to monitor the channel, this scheme outperforms other alternatives only if the energy consumption of the WUR is negligible compared to that of the main receiver. The block diagram of a WUR-enabled sensor device is shown in Fig. 1d.

A WUR is a simple receiver that receives the wake-up command (e.g., the device unique address) and generates an interrupt for the main receiver. In a WEH device, the implemented rectifier followed by a data slicer (comparator) can perform as a WUR with minimal complexity overhead. We advocate the use of such a WUR

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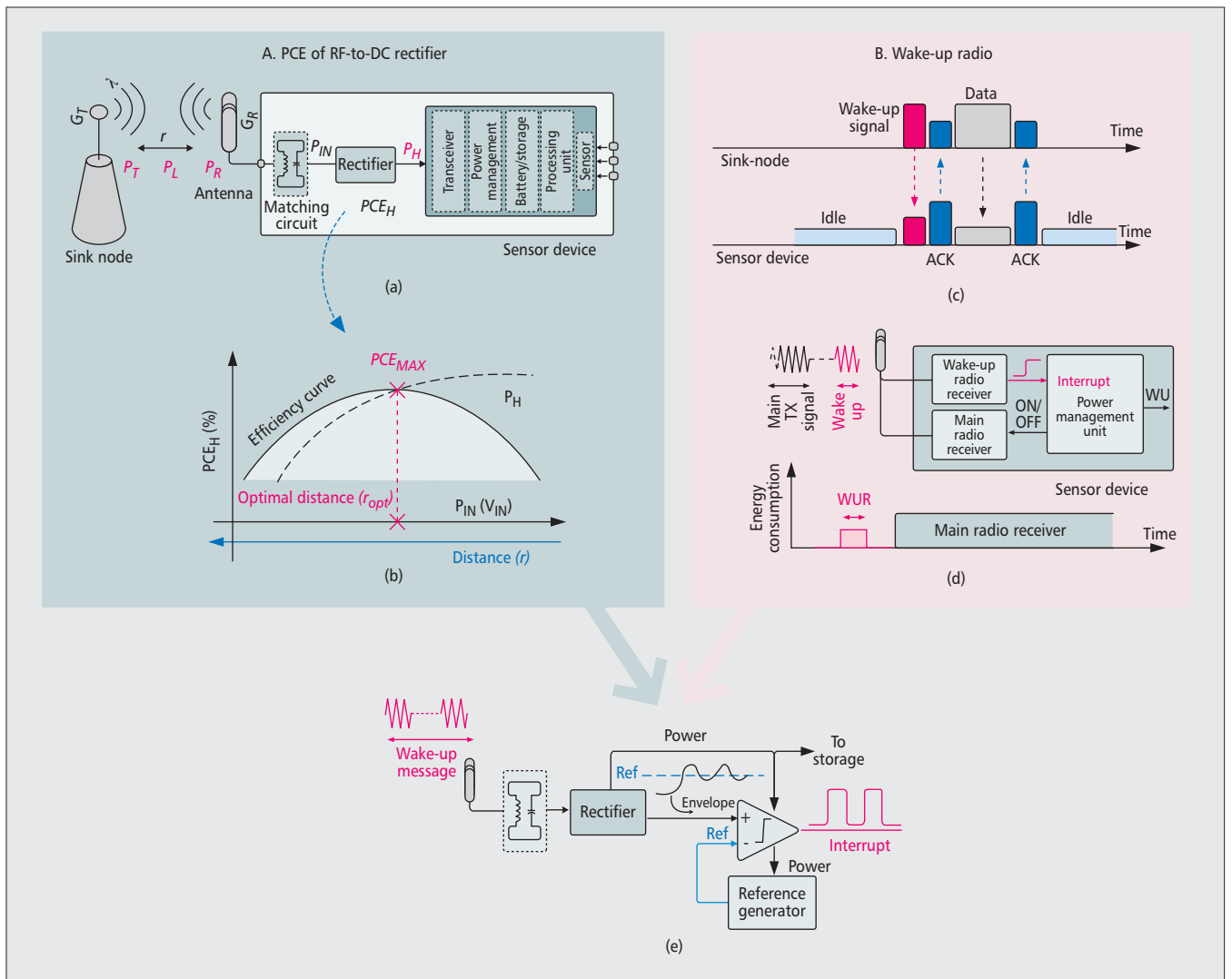


Figure 1. A WUR-enabled energy harvesting unit: a) block diagram of the WEH sensor device; b) efficiency curve (solid line) and harvested energy of the rectifier (dashed line); c) timing diagram of asynchronous wake-up scheme; d) block diagram and energy consumption of the WUR; e) zero-power interrupt generation unit.

(for detection of a simple on/off keying [OOK] wake-up message) as schematically shown in Fig. 1e [9]. If all the required energy of the WUR circuitry is supplied by the harvested energy from the wake-up message itself, the battery is not used during the idle listening mode (virtually zero energy consumption). Aside from energy consumption, high sensitivity and range, robustness to interferers, selectivity, and latency are also of paramount importance in designing a WUR for IoT.

POWER MANAGEMENT UNIT

An integral part of any energy harvesting system is its power management unit (PMU). The PMU is in charge of controlling the storage of the harvested energy. It also manages the distribution of the available energy among different consumers in an effort to maximize the lifetime of the device while maintaining high QoS. We extend the architecture of the PMU proposed in [10] to enable effective cooperation with the WEH unit. The architecture proposed is an

event-triggered/asynchronous scheme based on the signal generated by WUR. The PMU architecture also detects/preempts the failure of a node in the event of energy deficiency.

A detailed block diagram of the PMU for the WEH sensor device is shown in Fig. 2. The PMU starts its operation by a trigger signal generated by the WUR of the WEH unit (*INTERRUPT*). The PMU first activates the main transceiver through (*ON/OFF*) and then sends a wake up signal (*WAKE UP*) to the sensing unit to start its operation. The sensing unit toggles the *STOP/RUN* to high, signifying the PMU that it is in running mode. The *REQ* signal indicates the amount of energy required by the sensing unit. The signals *BAT* and *SE* indicate the amount of energy left in the battery device and the WEH unit storage element, respectively. Accordingly, the PMU activates switch *SW₁* through signal *SENSE* to fulfill the power requirements of the sensing unit. The sensor unit is in charge of sensing, data processing via a microprocessor (μ p) and finally transmitting them to a low-power transceiver based on Bluetooth, WiFi,

IEEE 802.15.4, Zigbee, and so on. The sensor device requires a minimum power of P_{Dmin} to operate in sensing mode. When the energy in the battery device goes below a certain threshold, $P_{TH} < 1.5 P_{Dmin}$, the PMU sends a *RECHARGE* command to the storage element by activating switch SW_2 of the WEH unit to charge the battery. When the energy level of the device remains $1.1 P_{Dmin}$, the device sends an out of service (OOS) command to the sink node, signaling that it is going out of service until it recharges itself again to more than $1.5 P_{Dmin}$. The sink node in turn sends a stop all service (SAS) signal to the device. The sink node/gateway puts the device out of the sensing service loop but keeps transmitting RF energy for harvesting. When the device is ready for service again, it sends a *READY* signal to the sink node, which in turn sends a resume all services (RAS) signal to the device.

LIFE-TIME PREDICTION THROUGH ENERGY COST MODEL

Let us consider a network of k static and identical sensor devices. As in [11], WSNs are either uniformly distributed in a ring topology, communicating with the sink node in a peer-to-peer fashion, or randomly distributed in a multihop ad hoc topology. As WSNs are a subset of an IoT system, we base our analysis on these topologies. Table 1 delineates the parameters used for the analysis of our scenarios.

The power transmitted by the sink node has a certain maximum and minimum value of P_{Tmax} and P_{Tmin} , respectively, and the sensor devices require a minimum power of P_{Dmin} to function properly. The sink node communicates with sensor devices asynchronously. We define two operational modes for the sensor devices: *active* and *idle* modes.

The lifetime of an IoT system (battery operated only/battery with WEH unit) depends on the average energy consumption of the sensor devices E_D per active duty cycle. This involves the combined operations of sensing, processing, and communication (receive/transmit).

Let N be the total number of active duty cycles representing the lifetime of the sensor device. The communication energy consists of E_{LS} (listening energy), E_{RX} (receiver energy), and E_{TX} (transmitter energy). The computation energy includes E_{PR} (processing energy) and E_{SN} (sensing energy). To capture the energy distribution among the aforementioned energy consumers, weighting coefficients $\alpha_{LS} > \alpha_{TX} > \alpha_{RX} > \alpha_{PR} > \alpha_{SN}$ are assigned to them. The total average energy consumption $E_D = \alpha_{LS} E_{LS} + \alpha_{TX} E_{TX} + \alpha_{RX} E_{RX} + \alpha_{PR} E_{PR} + \alpha_{SN} E_{SN}$. E_B is the total energy stored in the battery, and E_H is the available harvested energy per active duty cycle. We assume constant energy consumptions for receiver, processor, and sensor. However, the energy consumption of the transmitter (E_{TX}) is directly proportional to r_{ij}^2 , where r_{ij} is the distance between the originating device j and the sink node i (in ring topology) or the sink node/sensor device (in multihop topology). The harvested energy E_H is inversely proportional to

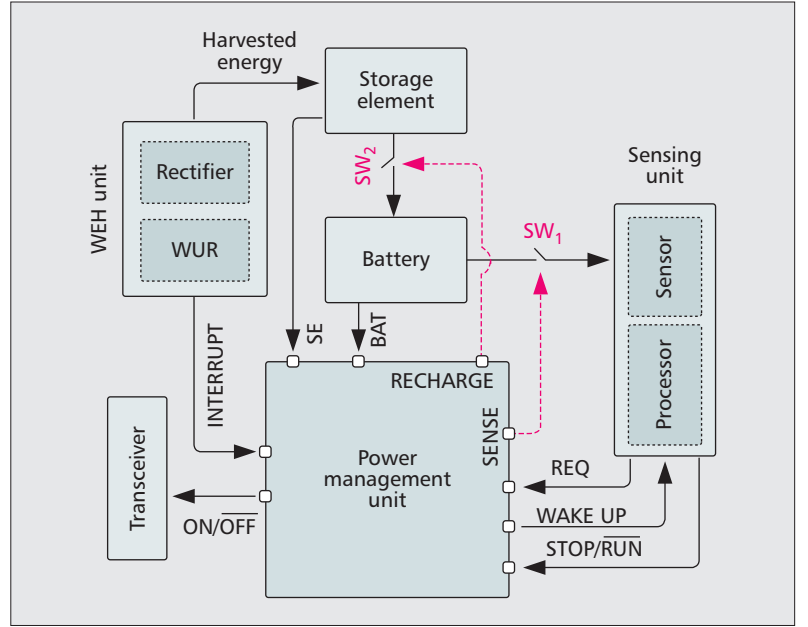


Figure 2. The proposed power management unit.

r_{ij}^2 (here j is the sink node and $r_{ij} = r_{ji}$). Based on these assumptions, the lifetime estimation N_H for the sensor device operated with a battery assisted by a WEH unit can be formulated as

$$N_H = \frac{E_B}{C_D + CTX \cdot r_{ij}^2 - \frac{C_H}{r_{ji}^2}} \quad (1)$$

where $C_D = \alpha_{LS} E_{LS} + \alpha_{RX} E_{RX} + \alpha_{PR} E_{PR} + \alpha_{SN} E_{SN}$. C_{TX} , and C_H are the proportionality constants for E_{TX} and E_H as mentioned above. We analyze the lifetime of the sensor devices for two different scenarios. Note that the lifetime of the battery-only device (without WEH) is denoted as N_B .

SCENARIO I: UNIFORM DISTRIBUTION IN A RING TOPOLOGY

In the ring topology, sensor devices (j s) are uniformly distributed around the sink node (i) at a distance r_{ij} , as shown in Fig. 3a. Since they are positioned equidistant from the sink node, all the devices receive a similar amount of wireless energy for harvesting. Assuming the channel is static between the sink node and the sensor devices, the transmitted energy from the device to the sink is the same for all nodes. Figure 3b schematically shows the distribution of energy among different consumers. The horizontal axis depicts the active duty cycle (T). Energy consumption for E_{TX} , E_{RX} , E_{PR} , and E_{SN} occur only for a fraction of the total active duty cycle, whereas energy consumption for E_{LS} and harvested energy E_H happen constantly throughout T . Based on these assumptions, the lifetime of the devices is estimated using Eq. 1 for N_B (battery only, E_H is zero) and N_H . Figure 3c shows the estimated lifetime vs. the power consumption of the devices (E_D). By incorporating the WUR scheme, we reduce the $\alpha_{LS} E_{LS}$ to have an almost negligible effect on C_D .

Parameter	Description
E_D	Total energy consumption of device
E_B	Energy of the battery
E_H	Harvested energy
E_{LS}	Listen mode energy
E_{RX}	Receive mode energy
E_{TX}	Transmit mode energy
E_{PR}	Processing energy
E_{SN}	Sensing energy
$\alpha_{LS} > \alpha_{TX} > \alpha_{RX} > \alpha_{PR} > \alpha_{SN}$	Weighting coefficients of the respective energy parameters
r_{ij}	Distance of originating device j to sink node i
N	Estimated device lifetime
N_B	Estimated device lifetime (battery only)
N_H	Estimated device lifetime (WEH-assisted battery)
$N_H(WUR, r)$	Estimated device lifetime (WUR-enabled WEH assisted battery) at distance $r = r_{ij}$

Table1. Parameters used for lifetime analysis.

Typical values for the distribution of different energy consumers are from [9, 12]. Energy harvesting increases the lifetime of the battery assisted device (N_H) by ~ 30 percent for low-power sensor devices (e.g., temperature, pressure, and light sensors). The WUR scheme enhances the lifetime by a further ~ 110 percent as depicted by $N_H(WUR, r)$. Setting $r_{ij} = r/2$, the lifetime of an energy harvesting device increases with the reduction in distance between sink node and sensor node as shown in $N_H(WUR, r/2)$.

SCENARIO II: RANDOMLY DISTRIBUTED MULTI-HOP TOPOLOGY

In the multihop topology, sensor devices (j_{NS}) are randomly distributed following a Poisson distribution. The maximum distance of the k th-farthest sensor device from the sink node is r_{ijk} . In a multihop transmission, the j_k sensor node's data hops $k - 1$ times before it reaches the sink node/gateway. The farthest node j_k only acts as a sensor and transmits its own data. However, the remaining j_1, j_2, \dots, j_k nodes act as both a sensor for their own data and a relay for data coming from farther nodes. Thus, to compute their lifetime, it is required to add the energy consumption of the relay cycle to that of the sensor cycle. The relay cycle energy consumption is $E_D - \alpha_{PR}E_{PR} - \alpha_{SN}E_{SN}$, which is the total energy consumption for a device operation minus the processing and sensing energy. Figure 4a depicts a two-hop scenario. Figure 4b shows the distribution of energy among different consumers. The transmit mode energy is smaller in this case than in Fig. 3b in the ring topology as the transmission distance is reduced to $r_{j_2j_1}$. Figure 4c shows the energy distribution for the sensor node j_1 operating as a sensor and as a relay for j_2 . The estimated lifetime of the sensor devices in the two-hop topology is shown in Fig. 4d for nodes j_2 and j_1 (lifetime for battery only, N_B , and for WEH assisted battery, N_H). As shown in the figure, energy harvesting extends the lifetime of the nodes. Node j_1 is exercised twice as often as j_2 , so it consumes more energy and has a shorter lifetime than node j_2 . The lifetime enhancement through WEH for a node at distance similar to that of the ring topology is approximately 5 times larger (~ 510 percent as compared to ~ 110 percent in ring topology) for low-power sensor devices. Therefore, for the N -hop topology, the lifetime of the devices is further enhanced.

FUTURE TRENDS AND RESEARCH CHALLENGES

The approaches presented in this work are not exhaustive. For the proposed system to become more practically viable, there are research challenges ahead that need to be addressed.

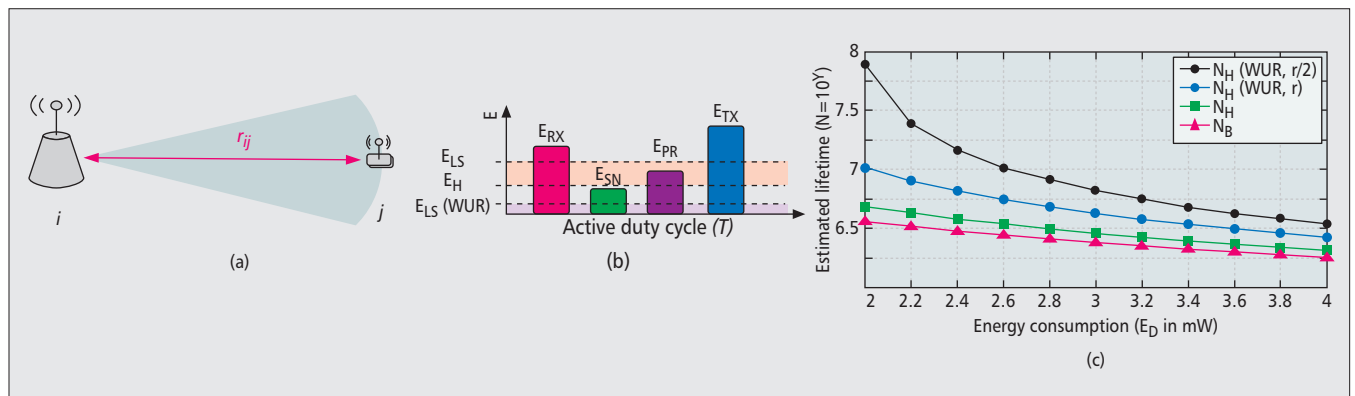


Figure 3. Lifetime prediction through energy cost model: a) ring topology architecture; b) energy distribution; c) estimated lifetime vs. energy consumption.

HIGHLY EFFICIENT, LOW-COST, AND SMALL-FORM-FACTOR WIRELESS ENERGY HARVESTING SYSTEM

The key challenge in successful large-scale deployment of sensor devices in an IoT infrastructure is to minimize their impact on users and the environment. Non-intrusive devices need to be small, be fabricated and deployed at very low cost, and are expected to operate in a self-sufficient manner for a long time. A WEH unit as an integral part of such devices must comply with such cost and size requirements. Efficiency is another crucial factor for a WEH system. High efficiency becomes increasingly relevant considering that the transmitted power by the dedicated source is usually limited due to health issues and interference constraints.

Commercial RF harvesting systems currently existing in the market enable single-band RF harvesting at sub-milliwatt power levels with efficiencies as high as 50 percent. However, extensive studies are still being carried out to improve the performance of WEH systems at the circuit and system levels. Energy beamforming [5], high gain antennas, and multi-band harvesting are among the other hot topics in the context of WEH systems for IoT.

CHANNEL STATISTICS FOR WIRELESS ENERGY HARVESTING IOT SYSTEMS

The scenarios and their respective analysis in our article assume the channel as static and time invariant. Practically, channel characteristics vary depending on the environment in which the number of interferers and the number of paths available from source device to sink. Harvested energy depends on the distance between sink and sensor node. In the presence of fading or multipath, the received energy for the purpose of harvesting and the transmitted data are adversely affected. In [13], a compressive sensing based approach is proposed to recover sparse signals from multiple spatially correlated data transmitted to a fusion center. Recently, in [14], researchers have proposed techniques to reduce the amount of packets to be retransmitted in case of faulty transmission, eventually saving energy.

CROSS-LAYER DESIGN OF WIRELESS ENERGY HARVESTING

Although the recent development of energy harvesting technologies mitigates the energy scarcity issue, the sensor device still has to operate in duty-cycled mode due to limited energy collection from the environment, and dynamically adjust duty cycles to adapt to the availability of environmental energy. Such dynamic duty cycles pose challenges for medium access control (MAC) layer protocol design in terms of synchronization, reliability, efficiency of utilizing channel resource and energy, and so on. Therefore, solutions of duty-cycling-aware middleware between MAC and physical layer power management are highly desired. Moreover, dynamic duty cycling also has nontrivial impact on the end-to-end performance of the network layer,

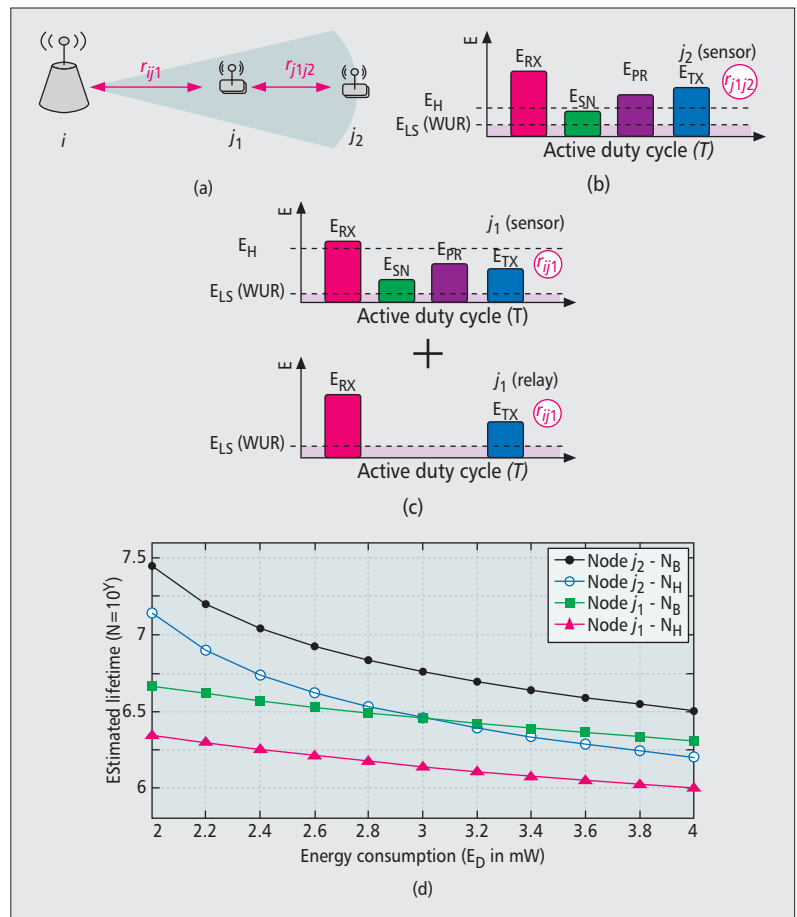


Figure 4. Life-time prediction through energy cost model: a) random distribution architecture; b) energy distribution of sensor j_2 ; c) energy distribution of sensor j_1 as sensor and relay; d) estimated lifetime vs. energy consumption.

including end-to-end delay, throughput, and so on. However, the current routing protocol design for IoT has paid very little attention to duty cycling. The problem of seamlessly integrating duty-cycle awareness into the multi-path routing scenario has been dealt with in [15] using a sleep scheduling mechanism; however, it still remains an open question.

SMARTPHONE RELAYS: WIRELESS ENERGY HARVESTING IN 5G SYSTEMS

Fueled by increases in traffic and data demand, mobile technology is moving toward the fifth generation (5G), where everything will be connected via the Internet and accessed through the cloud. Relay techniques utilized by 5G can benefit wireless/RF energy harvesting [6]. Also, smartphones, due to their mobility, can act as gateways/sink nodes, relaying data for sensors in personal area networks and as a source of RF energy for harvesting purposes.

CONCLUDING REMARKS

In this article, we have overviewed technologies and schemes to enable WEH for IoT systems. With an emphasis on improving the efficiency of the WEH unit and reducing the energy con-

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sumption of the devices, the lifetime of WEH-assisted battery-operated systems in an IoT architecture are analyzed for two different scenarios. A study of the specific energy requirements of IoT devices reveals that achieving self-sustainability requires improved design techniques at both the circuit and system levels.

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BIOGRAPHIES

POUYA KAMALINEJAD (pkamali@ece.ubc.ca) received his B.Sc. and M.Sc. degrees in electrical and computer engineering from the University of Tehran, Iran, in 2006 and 2008, respectively, and his Ph.D. degree in electrical and computer engineering from the University of British Columbia (UBC) in 2014. He is currently a post-doctoral fellow at the

University of British Columbia. His current interests include RF and low-power integrated circuit design, wireless energy harvesting for RFID tags and wireless sensor networks, and sensor interface design.

CHINMAYA MAHAPATRA (chinmaya@ece.ubc.ca) received his B.Tech. degree in electronics and communication engineering from N.I.T Rourkela, India, in 2009, and his M.A.Sc. degree in electrical and computer engineering from UBC in 2013. He is currently a Ph.D. student of electrical and computer engineering at the University of British Columbia. His current interests include the Internet of Things, body sensor area networks, embedded systems, sensor cloud, and smartphone energy optimization.

ZHENG GUO SHENG (z.sheng@sussex.ac.uk) is a lecturer at the School of Engineering and Informatics, University of Sussex, United Kingdom. He is also a visiting faculty member at UBC and the co-founder of WRNode. Previously, he was with UBC as a research associate, and with France Telecom Orange Labs as a senior researcher and project manager in M2M/IoT. He also worked as a research intern with the IBM T. J. Watson Research Center and U.S. Army Research Labs. Before joining Orange Labs, he received his Ph.D. and M.S. with distinction from Imperial College London in 2011 and 2007, respectively, and his B.Sc. from the University of Electronic Science and Technology of China (UESTC) in 2006. He has published more than 30 international conference and journal papers. He is also the recipient of the Auto21 TestDRIVE Competition Award 2014 and Orange Outstanding Researcher Award 2012. His current research interests cover IoT/M2M, cloud/edge computing, vehicular communications, and power line communication.

SHAHRIAR MIRABBASI (shahriar@ece.ubc.ca) received his B.Sc. degree in electrical engineering from Sharif University of Technology, Tehran, Iran, in 1990, and his M.A.Sc. and Ph.D. degrees in electrical and computer engineering from the University of Toronto, Ontario, Canada, in 1997 and 2002, respectively. Since August 2002, he has been with the Department of Electrical and Computer Engineering, UBC, where he is currently a professor. His current research interests include analog, mixed-signal, RF, and mmWave integrated circuit and system design with particular emphasis on communication, sensor interface, and biomedical applications.

VICTOR C. M. LEUNG [F] (vleung@ece.ubc.ca) is a professor and holder of the TELUS Mobility Research Chair in Advanced Telecommunications Engineering in the Department of Electrical and Computer Engineering at UBC, where he completed his B.A.Sc. and Ph.D. degrees in 1977 and 1981, respectively. He has been involved in telecommunications research with a focus on wireless networks and mobile systems for more than 30 years, which has resulted in more than 700 journal and conference papers co-authored with his students and collaborators, including several papers that won best paper awards. He is a Fellow of EIC and CAE. He was a Distinguished Lecturer of the IEEE Communications Society. He has served/is serving on the Editorial Boards of many journals. He has contributed to the organization committees and technical program committees of numerous conferences. He was the winner of an APEBC Gold Medal in 1977, an NSERC Postgraduate Scholarship for 1977–1981, an IEEE Vancouver Section Centennial Award in 2011, and a UBC Killam Research Prize in 2012.

GUAN YONG LIANG (eylguan@ntu.edu.sg) obtained his Ph.D. from the Imperial College of London, United Kingdom, and his B.Eng. with first class honors from the National University of Singapore. He is an associate professor at the School of Electrical and Electronic Engineering, Nanyang Technological University, Singapore. His research interests broadly include modulation, coding and signal processing for communication, and storage and information security systems.