Using Local and Global Knowledge in Wireless Sensor Networks

Christopher Gwilliams
Cardiff University

November 18, 2013

Contents

List of Figures

List of Tables

Abstract

To Write...

Chapter 1

Introduction

A wireless sensor network (WSN) consists of a collection of heterogeneous
nodes with sensing and, typically, wireless capabilities. These sensing nodes
can be extremely complex and powerful devices with the ability to sense
multiple phenomena simultaneously, or they can be simple motes that have
limited processing power and are tasked with sensing one thing in their en-
vironment.

Upon deployment, these nodes use their wireless capabilities to form links
with their neighbours, where a neighbour is any node that is within trans-
mission range. The way that nodes discover, and communicate with, their
neighbours is defined by their routing protocol. Routing protocols vary based
on the purpose of the WSN, the requirements of data transmission as well
as the characteristics of the nodes. Communication between nodes is expen-
sive and drains the available power faster than any other action that a node
performs. For example, if a WSN is deployed in a building with consistent
power available, then the routing protocol does not need to be adapted to
ensure the nodes maximise their battery life by transmitting as little as pos-
sible. However, not every WSN has unlimited resources at their disposal and
these protocols, as well as the underlying structure of the network, are used
to ensure the network is able to perform well for as long as possible.

Each WSN is different and each will have different constraints, a WSN
that monitors traffic along a busy road may experience memory limitations,
whereas a WSN that is deployed in the middle of a desert may experience
power issues. Typically, however, all WSNs do the same thing: sense one
or more characteristics of their environment and forward that data on to a
specified endpoint.

1.1 The Local Knowledge Problem

The majority of WSNs do not know what data they are sensing, or have any knowledge of their environment. This means that, unless fixed by a routing protocol or human that deployed the network, data is delivered on a chronological basis and is then filtered at the base station, usually manually. Some WSNs store all of the data on the node and users of the network must use a 'pull' model to query for data from nodes, but this requires some technical knowledge and, while it does increase the battery life of the nodes, it is a manual process again.

The environment that a WSN is deployed is usually rich and the data sensed often contains patterns that can be used to improve the performance of the network. For example, if a node knows that it is has only been triggering between the hours of 6pm and 5am for the past few weeks, it can enter a deep sleep outside of those hours or use that time to transmit data it has been storing while it knows it will be inactive. Alternatively, this knowledge can be used to prioritise data throughout the network so that the most important data is received first, instead of the most recent. An example of this could two camera nodes deployed facing the entry and exit of a building, tasked with looking for intruders between 5pm and 8am. If the camera facing the exit is triggered at 5:01pm and the camera on the entrance is triggered at 5:05pm, then the knowledge that the security guard leaves through the exit between 5:01pm and 5:08 pm will allow the entrance camera to prioritise its capture as more important, as it is an irregular occurrence.

This knowledge can be categorised into *local* and *global*. Local knowledge (LK) is the knowledge of an area that has been gained through experience or experimentation and global knowledge (GK) is knowledge that is generally available to everybody. An example of this is someone who has been tasked with deploying a WSN in the Amazon rainforest would use readily available sources, such as the Internet or prior research, to determine the humidity and weather patterns in order to use a node that could withstand such conditions. This would be classed as GK. However, a native to the Amazon may know that three of the locations that the nodes are to be deployed in are flooded for two weeks of the year, rendering their readings useless for that time period and increasing their risk of failure. This is LK, as it cannot be gained without experiencing the flooding in that area, or experimenting with water levels.

We believe that the use of this knowledge can increase the efficiency of the network, as well as prioritise sensed data by its value instead of the time it was recorded. To show this, we have developed a network architecture for WSNs that utilises knowledge from the data it senses, as well as its deployed environment. It is called the Knowledge-based Hierarchical Architecture for

Sensing (K-HAS) and this thesis will show how K-HAS addresses the problem of delivering the most important data first and improving the overall efficiency of the network.

1.2   Motivation

Throughout this thesis, we focus on a scenario motivated by our collaboration with Cardiff University School of Biosciences, who run a research centre in the Malaysian rainforest, in Sabah, known as Danau Girang (DG) . Located on the banks of the Kinabatangan river, DG has been running for more than six years and holds Masters, PhD and Undergraduate students from around the world, studying the ecology and biodiversity of the unique region.

The reason that the rainforest that DG is set in is so unique is that the area was heavily logged until the late 1970s and now serves as a corridor, between large Palm Oil plantations, connecting two separate rainforest lots. The area is now secondary rainforest (rainforest that has grown since being destroyed) and is experiencing a large variety of wildlife using the area as a habitat, or as a path. Some of this wildlife is unique to this area of the world and DG has had sightings of animals that have not been seen in many years.

There is a variety of research projects currently underway in the field centre, looking into fish population, crocodile attacks, hornbill habitats or the movement patterns of small mammals. One project that has been running almost since DG opened, is the *corridor monitoring programme*, a programme that consists of dozens of wildlife cameras deployed in various areas around DG and triggered whenever an animal triggers a break in their infrared (IR) sensor.

The Kinabatangan is a very humid place, with thick forest, making it very difficult to walk through and even more difficult for hardware to survive the conditions. Cameras are placed along the river and up to 1km into the forest, recording triggers onto SD cards. These SD cards are collected and stored at the field centre, where the images are manually collated and processed. The cameras are designed to have a battery life of three months but, due to the humidity, a battery life of 3 weeks is more realistic. In 2010, twenty cameras were deployed and half of them were inspected every two weeks, on a rotating basis. In that time, each cameras can record more than a thousand pictures and the dynamic nature of the rainforest, such as the sun through leaves, falling trees and reflections in the water can cause the camera to trigger when an animal has not walked past; we call these *false triggers*. False triggers can make up to 70

We have used this scenario to test our hypothesis and implement a WSN

that automates the collection, transmission, processing and storage of images, using LK to classify the data and prioritise the flow of information through the network, making more efficient use of the limited power and bandwidth available.

1.3   Research Contributions

1.4   Thesis Structure

The rest of this thesis is structured as follows. Chapter 2 provides some background on wireless sensor networks and the use of knowledge. Chapter 3 explains some technical decisions we made and the findings when running experiments in the Malaysian rainforest. Chapter 4 introduces the K-HAS architecture we have developed and explains the purpose of each tier. Chapter 5 details the ontology we have developed to support K-HAS and shows how current ontologies do not sufficiently cover all of the concepts involved with a *scientific observation*. Chapter 6 shows how we use rules to determine how valuable a piece of sensed data is and to explain how K-HAS is able to use these rules, and human feedback, in order to inform future classifications. Chapter 7 highlights the technical limitations of implementing K-HAS today and shows our simulations of it running as it was designed. Chapter 8 then concludes this thesis and summarises our contributions and findings, as well as highlighting work that could be undertaken to take this project further.

Chapter 2

Background

Using knowledge in a WSN is related to existing research into sensor networks that utilise context-awareness in order to improve their efficiency or adapt their sampling rate.

This chapter is split into the following sections. Section 2.1 outlines the issues surrounding WSN design and deployment. Section 2.2 details relevant existing routing protocols for sensor networks. Section 2.3 highlights popular sensor middleware in use today. Section 2.4 shows some examples of existing WSNs that are related to our motivating scenario. Section 2.5 introduces research into local and global knowledge and Section 2.6 shows some related work into WSNs that utilise knowledge or context to prioritise data and/or improve efficiency.

2.1 Wireless Sensor Network Issues

WSNs have been used in a number of domains, for a range of different purposes, from habitat monitoring [?] to military purposes [?] and healthcare [?]. While these applications are vastly different, the technology behind each is very similar. Each requires the use of nodes with sensors attached and each node requires a power source and storage devices.

According to [?], there are eight factors that affect the design of sensor networks, but we focus on a subset that are the most relevant to our research problem. The following points must be considered when designing a WSN:

2.1.1 Fault Tolerance

WSNs typically contain a large number of nodes and each can fail for various reasons, from a lack of power, filling its storage capacity, to factors of the

environment causing the hardware to fail. While the sensor nodes that are used in WSNs typically consist of the same platform, the variation between each deployment means that the device itself must be adapted to its environment, [?] used a custom protective casing for their nodes so that they were able to survive being in the open while ensuring that the transmission range was not affected.

2.1.2 Hardware Constraints

A sensor node typically consists of: a platform that contains the memory and processing power, a sensor (or sensors) and a transceiver that uses a wireless standard, such as Wi-Fi or Zigbee. Cost and size are the most common barriers to entry when designing a WSN. [?] mentions that the expectation of a sensor node is a matchbox-sized form factor. While the research is over ten years old, the original focus on sensor node was *smart dust* [?], small, inexpensive, disposable nodes that can transmit until their power reserve is depleted, [?] mentions that it is a requirement for the nodes to cost less than USD10. In [?], it is noted that, a decade on from the first WSN papers, smart dust has not been realised and the focus has instead been on larger, more powerful nodes that have reduced in cost and grown in power.

The Gartner Hype Cycle for 2013 [?] shows that smart dust is still in early innovation stages and may not be fully commercialised for another ten years. To counter this, research has been focussed on using software solutions to maximise the battery life in these more powerful, more expensive nodes, accompanies with the use of renewable energy sources.

2.1.3 Energy Constraints

The majority of sensor nodes do not have access to a constant power supply and must run on a battery that is, generally, a similar size to the node itself; or smaller. This means that the nodes must be as efficient as possible, knowing when to transmit data and when to sleep. The lifetime of a sensor network is extremely dependent on the battery life of each node and, unlike other mobile devices, they cannot typically be recharged [?]. Much work has been done on power efficient routing protocols, as well as the control of which attached devices are active [?, ?, ?].

The limited resources on the nodes mean that the sensing devices, and transceivers, attached must consume as little power as possible. Some routing protocols implement turning off wireless radios and scheduling a wakeup across the network [?], but the cost of turning off a device can waste just as much energy as leaving it on and sampling at a lower rate; if not more [?].

The use of energy in a node is dependent on how active the sensor(s) are, how much it transmits and receives, the transmission medium used as well as the environment it is in.

2.1.4 Transmission Medium

Common transmission media, such as Wi-Fi, are viable solutions in WSNs when a high data rate is required and power is readily available. However, research has shown that Wi-Fi is extremely power-hungry and [?] shows that Wi-Fi consumes almost 9 times more energy, while transmitting, than other standards, such as Zigbee. Bluetooth is a more power efficient standard that is becoming increasingly popular for sensing devices that are [art of the *Quantified Self* movement [?], with wearable devices that report measurements, such as heart rate, steps taken and calories burned. With the advent of the new low-power Bluetooh 4.0 , also known as Bluetooth Low Energy (BLE), this standard is supposed to allow months of continuous use on a coin-cell battery [?]. However, the range is limited to 100m and, using the same frequency, as Wi-Fi (2.4GHz) means that it is as susceptible to path loss and reduced transfer rates. [?] shows that a 2.4GHz Wi-Fi antenna is capable of transmitting up to 350m, while a considerably lower frequency of 41MHz was able to achieve links of 10km. The use of 2.4GHz frequencies in wet conditions have been shown to reduce the performance by up to 28New low-power, low-frequency standards have emerged in recent years and allow for a considerably longer range and increased battery life, at the cost of transmission speeds. Digmish is an example of this and, while it can achieve 250kb/s using 2.4GHz, it has much slower speeds of 125kbps when using the 900MHz spectrum. However, it does offer a range of, up to, 64Km [?].

2.1.5 Environment

The environment that a node is deployed in can have a great impact on almost all aspects of a WSN, such as range and battery life. Harsh environments that are not easily accessible make it difficult to place nodes and protected environments may limit where nodes can be placed. In [?], nodes were deployed within glaciers and had to survive extreme temperatures, lasting without human intervention, for months at a time. [?] attached collars to Zebras that had to withstand high speed movement, dust and high temperatures. The deployment of any node requires extensive research as to the environment that it will be deployed in and adjustments must be made to ensure it is able to survive for extended periods without continued maintenance. Section 2.1.4 also shows that environment does not simply affect the

hardware, but humidity can reduce the transmission range significantly, as well as moisture collecting on wireless antenna can reduce the range for days at a time.

2.2 Routing Protocols

Routing protocols specify how nodes in a WSN are organised, as well as how they transmit data throughout the network. In [?], the more popular routing protocols are surveyed and split into three of the main identified categories: data-centric, hierarchical and location-based. We use the aforementioned categories, as well as flat, to highlight some of the key protocols that are relevant to our work. The protocols have the task of ensuring that a network is performing at its best, providing the best lifetime and ensuring reliable and consistent delivery of data. This must reduce *flooding*, where nodes send every message to every link, aside from themselves, effectively flooding the network with unnecessary messages, and find a way to deliver data to the endpoint using the most efficient path possible.

2.2.1 Flat

Initially, this was the most common structure for a WSN, dozens of nodes spread out over a geographical area, with one or more neighbours, sending observations to a single endpoint.

MCFA

The Minimum Cost Forwarding Algorithm (MCFA) is an flat routing protocol that works by assigning costs to each node, based on how many hops they are from the base station [?].

Each node has a path-estimate of the cost of transmission from itself to the base station. The base station sends out a broadcast message and it is received by all nodes in range. The message contains a cost from the base station (initially zero) while every node has their cost set to infinity. If the cost in the message, plus the link it was received, is less than the current cost. If yes, the estimate is update on both the node and the message; the message is then passed on to other nodes in range.

This approach allows for dynamic reconfiguration of the network, as well as a reduced overhead due to not having to maintain a global routing table on each node. The assumption with MCFA, however, is that the direction of routing is always towards a fixed endpoint.

2.2.2 Data-centric

Data-centric routing protocols are not like traditional WSNs where nodes are given addresses; they use a method that involves the advertising, or querying, of the data that has been sensed and those with the relevant data can respond to the request.

SPIN

Sensor Protocols for Information via Negotation is one of the first data-centric protocols and attempts to address the issue of flooding the network whenever new data is sensed by addressing the data through metadata [**?**]. SPIN works on three messages passed between nodes:

1. ADV - A message sent by a node when it has sensed new data, advertising what it has recorded.

2. REQ - Sent by nodes that received an ADV to request the data.

3. DATA - Message containing the sensed data.

When a node has sensed data, it sends an ADV message to all nodes within range. If any of those nodes are interested in the data, then they respond with a REQ message, at which point the DATA message is sent to nodes that responded.

SPIN eliminates the need for a global view of the network topology, as nodes only need to know their single hop neighbours. However, SPIN does not guarantee equal diffusion of data throughout the network as a node may be interested in the data sensed at the other edge of the network, with only nodes that are not interested in between. This would mean that those nodes would not request the data or pass it on.

SPIN-IT

An extension to SPIN, SPIN-IT uses a slightly different approach to receiving data and is developed solely for the transfer of images [**?**].

Nodes use the existing message structure of SPIN, but REQ messages are used as queries, sent to all nodes in transmission range. The receiving nodes keep these requests and generate a new REQ message, thus allowing nodes to store temporal paths. When a REQ reaches a node that has the desired data, it responds with a ROUTE-REPLY message. This message is used because images are large and resource-constrained WSNs would have a much shorter lifetime if a lot of unnecessary transmissions were made. The

ROUTE-REPLY is used in case multiple nodes, in range of the requesting node, have the data it has requested and it can then choose the optimal route. As each node keeps a history of REQ messages, these can be used to trace the requested data back through the network, to the originating node, without the overhead of maintaining a global routing table.

COUGAR

A slightly different data-centric approach is the proposed COUGAR protocol, viewing the network as a distributed database. While similar to SPIN due to the fact that it does not forward data as soon as it is sensed, COUGAR uses a query language that abstracts the underlying network structure and uses that query to generate a plan that utilises in-network processing to provide an answer [**?**].

Within the network, a *leader* is selected and this node is used to aggregate the data from nodes that were able to fulfil all, or some, of the query. At risk of failure, each query should result in a *leader* being dynamically selected and it must have sufficient resources to be able to satisfy the request. This protocol was only proposed, and much of the technical detail has yet to be completed, but the concept of treating the network as a distributed database is a novel idea and this is one of the first protocols to suggest the use of a query language that could be used by people without specific domain knowledge.

2.2.3 Hierarchical

Hierarchical networks are WSNs that contain nodes of different classes; nodes at the end of the network are typically clustered into groups and served by a gateway node. This gateway could be in charge of aggregating the data, processing the data, or simply forwarding it to an endpoint. Clusters of nodes allow the network to be spread out over a wider geographical area and gateway nodes can use a different transmission method to provide long distance links to the base station. Gateway nodes serving a cluster of nodes means that the network can scale easily as well, simply by adding a new cluster to the network.

TEEN

The Threshold sensitive Energy Efficient sensor Network (TEEN) protocol is designed for reactive sensor networks, networks that require instant reactions to changes sensed in their environment. TEEN recognises that transmission is the most power hungry action for a node so each node is coded with a hard

and soft threshold. The hard threshold is a value that makes nodes transmit the reading to their cluster head. Similarly, the soft threshold is a small change in the value of the sensed attribute that causes further transmissions.

During the initialisation of the network, the base station sends information about the thresholds and sensing attributes to all cluster heads in the network; the cluster heads then forward this on to all nodes in their cluster. When a node senses data over the hard threshold, it transmits to the cluster and only transmits again when new sensed values are greater than the hard threshold and the difference between the current sensed value and the previous is greater than the soft threshold [?].

Clusters are assigned for a period of time and then new clusters are selected by the base station, at which point new attributes and thresholds are broadcast to all nodes. This kind of protocol allows the network to be dynamic after deployment and allows user input based on the data that has been sensed in the previous cluster times.

For example, a network could be tasked with sensing humidity in a rainforest but the thresholds have been set such that nodes are transmitting readings that are not of interest. A user can change these thresholds and they will be pushed out to the nodes at the time that the next clusters are chosen, without any need to visit the node or configure them individually.

2.2.4 Location-based

Instead of using the physical addresses of nodes, or the data they store, location-based protocols are based on the region that nodes are deployed in. **FLESH OUT**

Span

Span is a protocol where nodes are selected as *coordinators* based on their positions. A node can decide to be a coordinator based on the amount of energy it has and the number of neighbouring nodes it would benefit if they were able to use it as a bridge [?].

An example of this would be node B placed between node A and C. C and A are unable to communicate directly so, when node B wakes up, it decides whether it should become a coordinator. It knows that it has sufficient energy levels and it can provide connectivity for a previously disconnected area of the network, so it chooses to become a coordinator, staying awake and routing sensed data to other coordinators, which form the backbone of the network.

Results showed that using Span, in a system that transmits using 802.11, provides an network lifetime increase of more than a factor of 2 over networks that just use the 802.11 protocol.

GEAR

The Geographic Energy-Aware Routing protocol (GEAR) is similar to SPAN in that it makes routing choices based on both energy-awareness and location. Each node maintains an *estimated cost* and a *learning cost* of forwarding a packet through its neighbours. The estimated cost is calculated using the distance to the packet destination and the energy remaining on the node whereas the learning cost is the estimated cost that takes holes in the network into consideration [**?**].

GEAR is designed to perform in two phases: forwarding a packet towards a region and disseminating a packet within a region. When sending a packet towards a destination, GEAR either sends a packet on to the node in range that is closest to the destination or, if such a node does not exist, then a hole is identified. If a hole is identified then the node that minimises a cost is selected.

To disseminate a packet within a geographic area, uses algorithms based on the density of the network. Recursive geographic forwarding is typically used but this can result in an endless loop if the density of the network means that the region is unable to contact the destination. In that case, restrictive flooding is used.

Provide a concluding section for routing protocols here?

2.3   Sensor Middleware

Acting as a bridge between the hardware and the user, sensor middleware is software that abstracts the underlying network from the user and provides a means of accessing sensed data and administrating how the network performs. These middlewares must not be specific to a single network and provide support for as many different sensor nodes as possible. In [**?**], a middleware is said to provide standardised services to many applications and perform operations that make effective use of limited system resource.

In [**?**], a middleware should include four major components: programming abstraction, system services, runtime support and Quality of Service (QoS) mechanisms. In this section, we will discuss the challenges surrounding middlewares for WSNs and highlight some existing middleware that are particularly relevant to our research problem and/or motivating scenario.

2.3.1 Issues

WSNs present a range of new challenges to existing middleware, due to their resource constraints, deployment environments and more. However, there has been research into the key issues that must be addressed in order for middleware to be considered suitable. While there have been a number of surveys into these challenges [**?**, **?**, **?**], we will detail those that we believe to be most relevant to our work.

Resource Constraints

It is rare that nodes in a WSN would have a constant power source, unlimited memory and a casing that can survive a harsh environment without decaying. In order to ensure that the lifetime of nodes is mazimised, middleware needs to offer a power scheduling system that makes efficient use of the hardware on the node, typically turning off the radio at particular times.

Ideally, a middleware will be able to coordinate nodes through wireless communication, making efficient use of transmissions and dynamically modifying sleep schedules based on the power remaining.

Heterogeneity

Not every node in a network will have the same capabilities, manufacturer or hardware. WSN middleware needs to provide a standard interface to add data, regardless of the format it is recorded in. Some middlewares have been built for a specific set of hardware, however this homogeneity can provide increase the performance and efficiency of the network by only supporting one device.

Real-world Integration

WSNs are often tasked with recording phenomena that are time-crucial, so a sensor middleware should provide a real-time interface to the data that it has sensed. Ideally this data would be available outside of the network, though the use of an API.

Quality of Service

This issue is perhaps the most complex as QoS could apply to almost all aspects of the networks, such as efficiently using bandwidth, 100% uptime for nodes, guaranteed packet delivery or access to data stores. Some of these requirements are managed by the implementation of the routiong protocol,

the middleware should be able to monitor deployed nodes and report on their current status, as well as identify failures.

2.3.2 Existing Middlewares

In this section, we identify existing middlewares, explain how they address the issues highlighted in Section 2.3.1 and highlight how they relate to our research. While there are a lot of existing middlewares, our research did not show any that directly utilised the environment to make informed classifications. We did, however, find some applications that utilise context and rules to administrate the network.

FACTS

One such example is the FACTS middleware, an approach that uses a fact repository to coordinate nodes. Rules can then be implemented to process sensed data and fired when certain conditions are met [?]. More traditional sensor middlewares control the network and manage sensed data, this rule based approach allows for more flexibility, where rules can control the transmissions and process the data upon receipt.

Figure 2.1 shows the FACTS architecture, with the middleware holding the rulesets and a dstributed fact repository. Data within the network is stored as facts, providing a standard data format throughout the network and hardware abstraction. When new facts are received, ususally because of new sensing data, the rule engine checks to determine whether any rules should be fired. The ruleset definition language (RDL) is introduced here and each ruleset contains a group of relevant rules. Each rule is given a priority so that, if more than one rule is triggered by a fact, then the higher priority rules are fired first.

While FACTS itself does not utilise any local knowledge, the repository is used as a source for all previously sensed data and would prove as an excellent source of knowledge to assist with the classification of future readings. Also, the ability to add new rulesets, without technical knowledge of the hardware of each node, means that users of the network have the ability to add knowledge in the form of rules as they learn it.

ITA Sensor Fabric

The ITA Sensor Fabric is collaboration project between IBM, the US Army and the UK MoD. Sensor Fabric, or Fabric, is a two-way messaging bus and

set of middleware services connecting network assets to each other and users [?].

The core difference between the Fabric middleware and others is that not every node is sensing all of the time, sensor nodes are tasked when there is a requirement and they stop as soon as that task has been fulfilled. Similar to sinks in a traditional WSN, Fabric utilises Fabric nodes, which run the following three pieces of software:

1. Message Broker - Provides the communication infrastructure.

2. Fabric Registry - Holds information about the current deployment, such as all nodes deployed, all assets, routing information and tasks. Deployed in the form of a database.

3. Fabric Manager - The main service on the node to track the status of connected sensors, establish communication channels, provide a container for processing, plug-ins and to extends the capabilities of the Fabric.

Fabric runs on a Publish/Subscribe model, a sensing requirement is sent to a messaging broker as a subscription and this is distributed through all Fabric nodes and, thus, all sensor nodes. Sensor nodes then publish their data and the relevant data is sent to all applications that have subscribed to the data.

The plugin structure of Fabric makes it stand out from existing middlewares, allowing its functionality to be extended through web interfaces.

Because Fabric has been developed for military purposes that cross countries, policy enforcement has been implemented to restrict access to the granularity of sensed data but these access levels do not simply apply to a military context. Using our motivating scenario, researchers and professors should see animal images whereas the Sabah Wildlife Department should see images of hunters and people in the forest.

GSN

The Global Sensor Networks (GSN) is a middleware that has been developed to manage heterogeneous sensor networks and be suitable for those without any technical knowledge [?].

GSN provides hardware abstraction through the use of *virtual sensors*, a data stream that abstracts implementation details from the actual sensed data. A virtual sensor can be comprised of many streams and it can even consist of many virtual sensors.

Virtual sensors are described using XML, with tags that consist of metadata for the sensor, the structure of the incoming data stream, SQL queries for processing the incoming data and querying times. What makes GSN stand out is that virtual sensors do not have to be sensors deployed within your network, or even sensors at all, some examples of GSN show virtual sensors being added that read in data from the weather websites. This also means that the underlying structure of the network is irrelevant to GSN, as well as the physical locations of the nodes. Unlike some middlewares that have an expectation of how data will be routed, GSN is decoupled from the routing protocol, allowing them to act independently.

GSN is completely open source and, while it does not provide the plug-in architecture that is available in Fabric, the Java code can be modified to suit a specific deployment.

Figure 2.2 outlines the architecture of GSN, showing that the virtual sensors are stored on a central node and their inputs are managed and stored. GSN also comes bundled with a web interface to show all active sensors and their most recent recordings, as well as the implementation of web services to access the data outside of the interface.

Data from virtual sensors pass through the virtual sensor manager to the storage layer. Once the data has been stored, the query manager is invoked and queries are loaded from the repository and executed by the manager. The results of the queries are then handled by the notification manager and also made available to the web interface. Notifications can be extended to support many different forms of communication, such as SMS, email or web services.

Virtual sensors do not natively support all hardware, although new virtual sensors can be described using XML, and there may be a need to implement an entirely new virtual sensor. In this case, technical knowledge is required, and new sensors can be implemented through use of the Java programming language. This provides more control over the use of XML and allows users to specify how sensed data is stored in a database, use external libraries to receive proprietary data, specify processing workflows before the data is stored or implement new notification methods for the users of the network.

To show the simplicity of a basic virtual sensor, [?] describes a temperature sensor that we reproduce here in Listing ??. The file is human-readable and has a shallower learning curve than programming languages, with tags that explain what data they contain. In this example, the output structure shows that only a temperature reading is received and that the data should be stored permanently. The stream source specifies the content of the stream and the query details the standard query that should be used to extract data from GSN.

Figure 2.1: FACTS Architecture

Figure 2.2: GSN Architecture
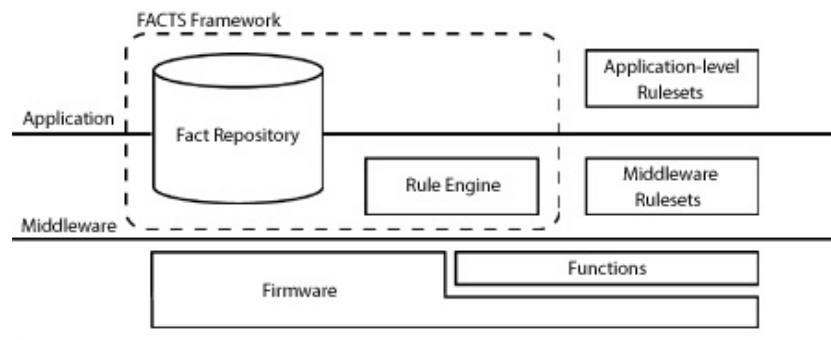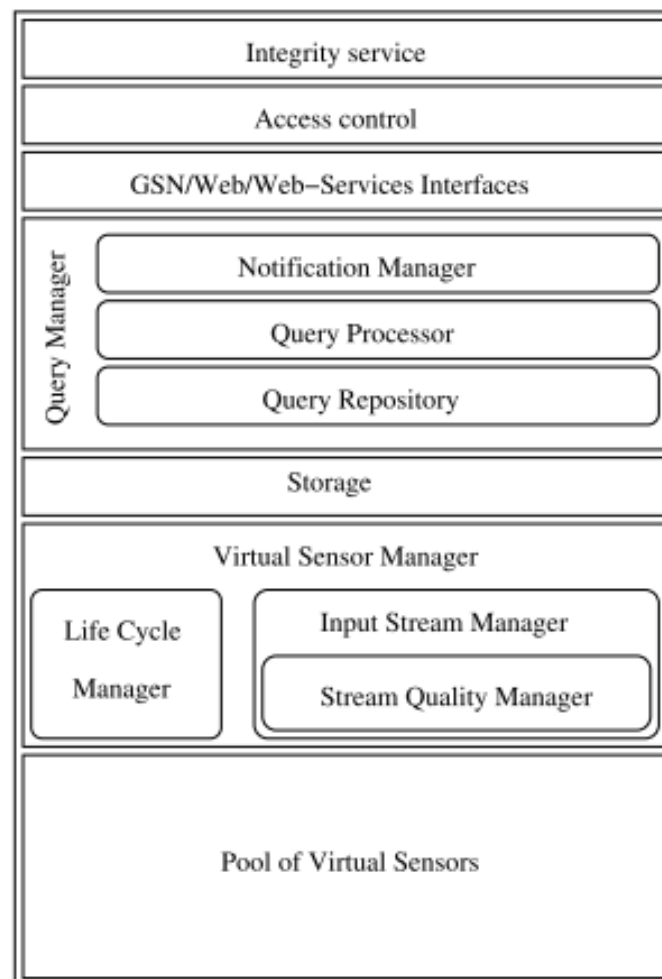
```
<life−cycle  pool−size="10"  />
<output−structure>
          <field  name="TEMPERATURE"  type="integer"/>
</output−structure>
<storage  permanent−storage="true"  size="10s"  />
<input−stream  name="dummy"  rate="100"  >
<stream−source  alias="src1"  sampling−rate="1"  storage−size="1h">
<address  wrapper="remote">  <predicate  key="type"  val="temperature"  />
<query>select  avg(temperature)  from  WRAPPER</query>
</stream−source>
<query>select  ∗  from  src1</query>
</input−stream>
\caption{Virtual  Temperature  Sensor}
\label{bg:lst:gsn}
```

The modularity and flexibility of GSN makes it different to existing middlewares as it has not been designed for any specific hardware and modules of the middleware can be replaced, such as the database.

2.4   Biodiversity Sensor Networks

In this section we will cover existing WSNs that are related to our motivating scenario or, more specifically, biodiversity focussed WSNs that have been deployed to monitor wildlife and/or the environment. WSNs for habitat, and wildlife, monitoring are especially important because these are areas that often need to be untouched by humans. Areas with high human disturbance can influence the abundance of species and some habitats, i.e. underground burrows, may be impossible to monitor without destruction.

One of the most well known WSNs to monitor habitat is the network deployed on Great Duck Island (GDI), an island off the coast of Maine, USA. A network consisting of 32 nodes was deployed to monitor a bird, known as the Leachs Storm Petrel [**?**]. This network used a clustering approach for groups of nodes to send data to a gateway, which would then route it back to the base station. The base station, located a few kilometres away on the island, has internet access and uploads the data to allow users to browse and process the data.

A multihop approach was used here as they found that, for sufficient coverage, single hop connectivity would not cover all of the island. Acrylic enclosures were developed to ensure the nodes were weatherproofed for the conditions of the island, while maintaining the functionality of each sensor and not impeding transmission range. While the nodes, their casing and

their sensors have been designed specifically for the deployment on GDI, the success of the network, running for 123 days in the early stages of WSN research [?], shows that this approach can be used elsewhere with similar effects; allowing hard to monitor and/or inaccessible areas to be continuously monitored.

On a smaller scale, INternet-Sensor InteGration for HabitaT monitoring (INSIGHT) is a single-hop WSN that allows remote access for data and reconfiguring of nodes [?]. While this network does use commercial hardware, their findings do show that their nodes could survive for 160 days on a single battery, supporting their claim that a single hop network allows for a longer network lifetime.

The key feature of this network is the ability for humans to remotely set reporting thresholds for sensor nodes. This means a user can prolong the lifetime of nodes by limiting the threshold they report on, as well as the fact that these thresholds are a way for users to add knowledge, albeit primitive, into a network.

While there is research on cameras used to monitor animals [?, ?], these networks are generally cameras deployed with their memory cards manually retrieved and processed. In recent years, however, the use of wireless technologies and image-based WSNs has increased, [?] uses wireless cameras to monitor the movement of animals between roads. Using commercial hardware and controlled sleep scheduling, this solution employs the use of nodes to detect movement and wake up more power-hungry camera nodes. While the nodes are wireless, the distance of the network from civilisation means that the data does still need to be collected manually and uploaded to a computer.

Due to the advent of smartphones and tablets, as well as the improvements in 3G technology, projects taking advantage of more modern technologies have grown in popularity. Using 3G enabled cameras, [?] have deployed a number of devices in locations all over the world, such as: Kenya, Indonesia and the USA. The images captured are transmitted to a server and a website allows the general public to not only see the images in near real-time, but to classify the images as well. This crowdsourcing of collective knowledge lets people, that may not have domain knowledge, vote on an image and those votes are used to make classification easier.

Over the past fifteen years, WSNs have grown from a concept to a real solution for monitoring the habitats, movements and eating habits of wildlife all over the world. Whether it is using GPS collars to monitor the movement of cattle [?], monitoring animal habitats on a remote island or using cameras to capture the animals themselves, the popularity of these networks has grown considerably and advances in technology have allowed these networks

to be deployed in places that humans cannot.

2.5   Local and Global Knowledge

The environment of a sensor network is rich and varied and we believe that patterns in the data sensed can be used to inform the network on decisions surrounding the transmission and processing of newly sensed data. As our research began, we simply called this knowledge but, as our work continued, it became apparent that it could be split further.

While we believe that we are the first to use the concept of local and global knowledge within the wireless sensor network domain, the terms have been around for many years. In 1999, a book that referred to local knowledge as *indigenous knowledge* defined local knowledge as systematic information that remains in the informal sector, usually unwritten and preserved in oral traditions rather than text [**?**].

Over the past twenty years, local knowledge has been used in various contexts, from researching lending and the credit market [**?**] to extracting local knowledge from natives to improve farming techniques [**?**]. This research, as well as work that will be covered later, showed us that there are two kinds of knowledge: global and local.

It was from agriculture research that we were able to refine our definition of local knowledge, [**?**] defines local knowledge as knowledge that farmers have derived locally through experience and experimentation. They also say that indigenous knowledge is different in that it is culturally specific. From this definition, as well as our work with our motivating scenario, we were able to generalise the definition and expand upon it.

We now define local knowledge as *knowledge of an area, held by a domain expert, that has been gained through experience or experimentation.* This then means that global knowledge is *knowledge of an area that can be accessed by anyone, without the need to visit the area directly.* The weather of a region is global knowledge because it can be found through a variety of media, whereas the level of a rain for a field within that region would be local knowledge, as it would require experimentation.

Using these definitions, we believe that encoding local and/or global knowledge onto sensors will inform routing decisions to make better use of the bandwidth in resource-constrained WSNs by sending data that it believes to be important first, rather than just chronologically. Patterns in the data, and knowledge of the environment surrounding a node, will allow a node to infer what the data may be classified as, automate the classification process, learn from previously sensed data and utilise global knowledge of ongoing

projects within the network to determine what data is thought to be of a higher priority.

2.6 Relevant Existing Networks

In this section, we discuss existing networks that have done work relating to our research question and/or our motivating scenario. While most of the existing networks relevant to Danau Girang have already been covered in Section 2.4, there are some WSNs that are not directly related to biodiversity but have been deployed in harsh conditions or involve interdisciplinary collaboration.

2.6.1 Context-Awareness

While standard WSNs have been prevalent for many decades, new research on the Internet of Things [?] has brought about interest in context-aware sensing, in some cases this is for small wearable devices to track fitness but the applications are much broader. Here we will look at WSNs that use context to make informed routing decisions, save power, or prioritise the transmission of data.

Health

Health monitoring is one of the more obvious choices for context awareness as classifying readings can often help determine the health of someone, rather than their self-reports. AlarmNet is a WSN that uses context to provide long-term health monitoring for people in assisted-living environments.

AlarmNet employs context-awareness to learn about the activity levels of the patient and uses that knowledge to determine when changes in the readings may mean that the patient is at risk. Once the AlarmNet system has completed its learning period, deviations, from what it has recorded as the normal activities for the patient, are sent to nurses and doctors, with the idea that this information can assist with a diagnosis.

As some nodes in the system will be battery powered, AlarmNet also employs a subsystem, called the Context-Aware Power Management System (CAPM), to control the power consumption of a device based on the recorded activities of the patient. The system uses policies, based on the context, to save power for all nodes in the system. For example, the system could put mains-powered nodes into a low-power state and disable all nodes outside of the bedroom when it detects that the patient has gone to sleep.

This use of context has not only shown that it can assist with the lifetime of a network, but it can also provide valuable insight into sensed data to provide data enriched with semantics.

Wearable Devices

As wearable devices, such as the Fitbit and the Pebble smartwatch, are increasing in popularity, context-awareness is a useful tool to differentiate between the many activities that a person can undertake. The MObile PErsonal Trainer system (MOPET) is a wearable fitness devices that uses context to record data on jogging and fitness exercises [**?**].

While the project is five years old now, it is one of the earliest proof-of-concept devices created and shows how simple rules applied to sensor readings can be used to apply context. For example, the device consists of a GPS sensor and, when active, it is constantly recording positions. Using pairs of points, it is able to calculate the speed and, if the speed is consistent with jogging, then it records a running exercise.

Fitness is not the only purpose for wearable computing, the eWatch is an early design for a *smartwatch* that uses a microphone and light sensor to record the locations that the user visits, storing previous recordings in flash memory and matching them with current recordings [**?**].

UNETS

*UNETS is tiered and context aware

2.6.2   Harsh Environments

There are a number of situations where we may want to record data in an environment that is not safe for humans, such as a volcano [**?**], or that may be difficult for electronics to survive. In these cases, special considerations must be made during the design and deployment of the WSN, in order to ensure maximum network lifetime with reliable readings.

GLACSWEB

Glacsweb is a sensor network to monitor the rate that which glaciers are melting. Deployed in Norway, specially designed sensors have been drilled into glaciers to monitor pressure, temperature, orientation and strain. Due to the high pressure and exposure to moisture, a polyester casing was used so that, once bonded, the node inside would not be recoverable.

Reventador

Volcano Reventador is in northern Ecuador and a WSN has been deployed on there to monitor eruptions. Similar to Glacsweb, weatherproof enclosures were used to prevent ash and moisture from breaking the sensors and long-range external antenna mounted to a pole was used to achieve communications over large distances when wireless communications have proven to be difficult [?].

Rainforest

There have been many studies on how the rainforest affects the range and quality of wireless links [?, ?, ?]. In [?], the humidity was shown to reduce 802.11 range by up to 78% and [?] explains that periods of rainfall reduce the link quality up to 100% in some cases, resulting in the loss of a hop and this could prevent data from reaching the endpoint.

2.7 Summary

In this chapter we have explored the components that make a WSN, as well as some existing deployments that are relevant to our motivating scenario. Sensor middlewares have come a long way in the past decade and increased capabilities for sensor nodes have allowed for more intense processing to be carried out before any data is seen by a human. Context is now used on sensor nodes to infer the activity it is undertaking or even to determine when it should wake to sample.

Local knowledge is not a new concept and it has been used for many years to extract information from indigenous people and in industry to adapt their processes to a local area, such as farming. However, we believe that our work is the first to apply local knowledge in the context of WSNs, but context-aware sensor networks support our hypothesis that knowledge of its surroundings, or of the data it has sensed, can improve network performance.

The two primary components of a WSN that could be injected with local knowledge is the middleware and the routing protocol, each providing different benefits. Existing work has shown middleware that uses context-awareness can use that to make global changes to the network, such as power management, whereas routing protocols affect the data that is sampled and sent to the endpoint(s) of the network, such as adaptive thresholds.

One issue in proving this hypothesis is the deployment of a network that utilises local knowledge. The deployment environment of our motivating scenario is not only interdisciplinary, it is in a region that is humid, dynamic

and dense. Existing research has shown that the deployment of a WSN in these considerations means that range will be greatly reduced and changes in humidity can prevent communication altogether, as well as moisture affecting the hardware itself. To deploy a network, hardware must be adapted and the right communication medium must be chosen in order to maximise link quality and minimise dropped connections.

In Chapter 2, we show how we used this research, as well as our own findings, to choose suitable hardware and experiments, that were informed by previous rainforest range tests in the literature, supported our choice for communication medium.

Chapter 3

Technical

In this chapter, we explain our motivating scenario in more detail, explore the sensor hardware that we researched and outline the results of experiments we undertook in the Malaysian rainforest. As highlighted in Section 1.2, we have been working with Cardiff University School of Bioscience to design and deploy a WSN that utilises local knowledge, using an area of rainforest in Malaysia owned by the Sabah Wildlife Department.

Bibliography

[1] K. Aberer, M. Hauswirth, and A. Salehi. Zero-programming sensor network deployment. *Next Generation Service Platforms for Future Mobile Systems (SPMS)*, 23:21–24, 2007.

[2] Karl Aberer, Manfred Hauswirth, and Ali Salehi. A middleware for fast and flexible sensor network deployment. In *Proceedings of the 32nd international conference on Very large data bases*, pages 1199–1202. VLDB Endowment, 2006.

[3] Jorge a Ahumada, Carlos E F Silva, Krisna Gajapersad, Chris Hallam, Johanna Hurtado, Emanuel Martin, Alex McWilliam, Badru Mugerwa, Tim O'Brien, Francesco Rovero, Douglas Sheil, Wilson R Spironello, Nurul Winarni, and Sandy J Andelman. Community structure and diversity of tropical forest mammals: data from a global camera trap network. *Philosophical transactions of the Royal Society of London. Series B, Biological sciences*, 366(1578):2703–11, September 2011.

[4] a.K Rahman, M.S Anuar, S.a Aljunid, M.N Junita, a. Sahadah, M.N Ismail, and B.S. Naimullah. Study of rain attenuation with diversity of wavelengths propagation transmission in tropical rainforest region. *2008 International Conference on Electronic Design*, pages 1–4, December 2008.

[5] Kemal Akkaya and Mohamed Younis. A survey on routing protocols for wireless sensor networks. *Ad Hoc Networks*, 3(3):325–349, May 2005.

[6] I Akyildiz. Wireless sensor networks: a survey. *Computer Networks*, 38(4):393–422, March 2002.

[7] IF Akyildiz. A Survey on Sensor Networks. *Communications . . .*, (August):102–114, 2002.

[8] Luigi Atzori, Antonio Iera, and Giacomo Morabito. The Internet of Things: A survey. *Computer Networks*, 54(15):2787–2805, October 2010.

26

[9] Daryoush Bayat, Daryoush Habibi, and Iftekhar Ahmad. Development of a Wireless Sensor Node for Environmental Monitoring. ... *International Conference on Sensor* ... , (c):1–5, 2012.

[10] Fabio Buttussi and Luca Chittaro. MOPET: a context-aware and user-adaptive wearable system for fitness training. *Artificial intelligence in medicine*, 42(2):153–63, February 2008.

[11] A. Chen. A scalable solution to minimum cost forwarding in large sensor networks. In *Proceedings Tenth International Conference on Computer Communications and Networks (Cat. No.01EX495)*, pages 304–309. IEEE, 2001.

[12] Benjie Chen, Kyle Jamieson, Hari Balakrishnan, and Robert Morris. Span: An energy-efficient coordination algorithm for topology maintenance in ad hoc wireless networks. *Wireless Networks*, 8(5):481–494, 2002.

[13] Peter Corke, Tim Wark, and Raja Jurdak. Environmental wireless sensor networks. *Proceedings of the IEEE*, 98(11), 2010.

[14] M. Demirbas. INSIGHT: Internet-Sensor Integration for Habitat Monitoring. *2006 International Symposium on a World of Wireless, Mobile and Multimedia Networks(WoWMoM'06)*, pages 553–558.

[15] B. R. DEWALT. Using indigenous knowledge to improve agriculture and natural resource management. *Human organization*, 53(2):123–131.

[16] D Estrin and L Girod. INSTRUMENTING THE WORLD WITH WIRELESS SENSOR NETWORKS. *Acoustics, Speech, and* ... , 2001.

[17] Carlos M. S. Figueiredo, Eduardo F. Nakamura, Afonso D. Ribas, Thales R. B. de Souza, and Raimundo S. Barreto. Assessing the communication performance of wireless sensor networks in rainforests. *2009 2nd IFIP Wireless Days (WD)*, pages 1–6, December 2009.

[18] Antonio-Javier Garcia-Sanchez, Felipe Garcia-Sanchez, Fernando Losilla, Pawel Kulakowski, Joan Garcia-Haro, Alejandro Rodríguez, José-Vicente López-Bao, and Francisco Palomares. Wireless Sensor Network deployment for monitoring wildlife passages. *Sensors (Basel, Switzerland)*, 10(8):7236–62, January 2010.

[19] S. Hadim and N. Mohamed. Middleware: Middleware Challenges and Approaches for Wireless Sensor Networks. *IEEE Distributed Systems Online*, 7(3):1–1, March 2006.

27

[20] WR Heinzelman, J Kulik, and H Balakrishnan. Adaptive protocols for information dissemination in wireless sensor networks. *. . . of the 5th annual ACM/IEEE . . .* , 1999.

[21] Mark Hempstead, Nikhil Tripathi, Patrick Mauro, Gu-Yeon Wei, and David Brooks. An Ultra Low Power System Architecture for Sensor Network Applications. *ACM SIGARCH Computer Architecture News*, 33(2):208–219, May 2005.

[22] C Intanagonwiwat, R Govindan, and Deborah Estrin. Directed diffusion: a scalable and robust communication paradigm for sensor networks. *Proceedings of the 6th annual . . .* , pages 56–67, 2000.

[23] L Joshi and M Van Noordwijk. BRINGING LOCAL KNOWLEDGE INTO PERSPECTIVE A CASE OF SUSTAINABLE TECHNOLOGY DEVELOPMENT IN JUNGLE RUBBER AGROFORESTS IN JAMBI. *Knowledge Creation Diffusion Utilization*, 2001.

[24] Philo Juang, Hidekazu Oki, Yong Wang Ý, Margaret Martonosi, Li-shiuan Peh, Daniel Rubenstein , Óññùò ØÓò, Ú Ó Ñ Ò Ö Ò ý, ÑôóöøÒø, Ö Ú Ö Ó Ñ Ò Ý Ò Û Óñôùø, Ò Ô Ô ØÓò, Ø Ó Ò Òúóú, Ò Ùøóòóñóù Ù Ó Óñôùø, Ò Û Ö Óññùò ØÓò, Ú Ó Ö Óø, Ôóý Ò Û Ö Ô Öøóô Ö Ò Øûóö Ò ØÒ Õ Ù Ò, Ñ Ó Ò Ó Ö Ò Øûóö, Ö ÆØÝ ØÑ Ò Ù Ù Øóñ, Ø Ö Ò Ó, Ó Ö Û Ö Øö, Ò Ú Ö Ò Ñ Èí, Ò Øý, Ó Ú Ö Ò ØÖ Ó Ò Û Ö Ò Ñ Ö Øù, Ó Ñ Ò Òøö, Ôöóøó Ó, Ò Ò Ö Ý ØÖ Ó Ôö, Ò Ó ØÖ Û Ö Ò Ò Óö, Ô Ô ØÓò, Ò ØÒ Õ Ù óû, and Ø Ñ ØÓ Û Ô Ò Øóö. Energy-Efficient Computing for Wildlife Tracking : Design Tradeoffs and Early Experiences with ZebraNet. *Structure*, 2002.

[25] JM Kahn, RH Katz, and KSJ Pister. Next century challenges: mobile networking for Smart Dust. *. . . on Mobile computing and networking*, pages 271–278, 1999.

[26] Roland Kays, Bart Kranstauber, Patrick Jansen, Chris Carbone, Marcus Rowcliffe, Tony Fountain, and Sameer Tilak. Camera traps as sensor networks for monitoring animal communities. *2009 IEEE 34th Conference on Local Computer Networks*, pages 811–818, October 2009.

[27] Jin-Shyan Lee, Yu-Wei Su, and Chung-Chou Shen. A Comparative Study of Wireless Protocols: Bluetooth, UWB, ZigBee, and Wi-Fi. *IECON 2007 - 33rd Annual Conference of the IEEE Industrial Electronics Society*, pages 46–51, 2007.

[28] Alan Mainwaring, David Culler, Joseph Polastre, Robert Szewczyk, and John Anderson. Wireless sensor networks for habitat monitoring. *Proceedings of the 1st ACM international workshop on Wireless sensor networks and applications - WSNA '02*, page 88, 2002.

[29] a. Manjeshwar and D.P. Agrawal. TEEN: a routing protocol for enhanced efficiency in wireless sensor networks. *Proceedings 15th International Parallel and Distributed Processing Symposium. IPDPS 2001*, 00(C):2009–2015, 2001.

[30] K. Martinez, R. Ong, and J. Hart. Glacsweb: a sensor network for hostile environments. *2004 First Annual IEEE Communications Society Conference on Sensor and Ad Hoc Communications and Networks, 2004. IEEE SECON 2004.*, pages 81–87, 2004.

[31] U. Maurer, a. Rowe, a. Smailagic, and D.P. Siewiorek. eWatch: A Wearable Sensor and Notification Platform. *International Workshop on Wearable and Implantable Body Sensor Networks (BSN'06)*, pages 142–145.

[32] Chris Otto and A Milenkovic. System architecture of a wireless body area sensor network for ubiquitous health monitoring. *Journal of Mobile . . .*, 1(4):307–326, 2006.

[33] Diego Pizzocaro, Fangfei Chen, Thomas La Porta, and Matthew P Johnson. System Architectures for Multi-Sensor Task Allocation. *Network.*

[34] Atiqur Rahman. Middleware for wireless sensor networks : Challenges and Approaches Sensor Network application : 3 Middleware for Sensor Network : Network :. *Security*, pages 2–6.

[35] C. Schurgers and M.B. Srivastava. Energy efficient routing in wireless sensor networks. *2001 MILCOM Proceedings Communications for Network-Centric Operations: Creating the Information Force (Cat. No.01CH37277)*, pages 357–361.

[36] Michael Segal. Improving lifetime of wireless sensor networks. *Network Protocols and Algorithms*, 1(2):48–60, 2010.

[37] Ladislaus M. Semali. *What Is Indigenous Knowledge?* Routledge, 1999.

[38] Joseph E. Stiglitz. Peer Monitoring and Credit Markets. *The World Bank Economic Review*, 4(3):351–366, September 1990.

[39] Melanie Swan. Sensor Mania! The Internet of Things, Wearable Computing, Objective Metrics, and the Quantified Self 2.0. *Journal of Sensor and Actuator Networks*, 1(3):217–253, November 2012.

[40] R. Szewczyk, Eric Osterweil, Joseph Polastre, Michael Hamilton, Alan Mainwaring, and Deborah Estrin. Habitat Monitoring With Sensor Networks. *Communications of the ACM*, 47(6):34–40, 2004.

[41] Robert Szewczyk, Joseph Polastre, and Alan Mainwaring. Lessons from a sensor network expedition. *Wireless Sensor*, pages 307–322, 2004.

[42] Kirsten Terfloth and Georg Wittenburg. Facts-a rule-based middleware architecture for wireless sensor networks. *Software and Middleware (*, 2006.

[43] N.H. Vaidya. A wakeup scheme for sensor networks: achieving balance between energy saving and end-to-end delay. *Proceedings. RTAS 2004. 10th IEEE Real-Time and Embedded Technology and Applications Symposium, 2004.*, pages 19–26, 2004.

[44] M.M. Wang, J.N. Cao, J. Li, and S.K. Dasi. Middleware for wireless sensor networks: A survey. *Journal of computer science and technology*, 23(3):305–326, 2008.

[45] Tim Wark, Wen Hu, Peter Corke, Jonathan Hodge, Aila Keto, Ben Mackey, Glenn Foley, Pavan Sikka, and Michael Brunig. Springbrook: Challenges in developing a long-term, rainforest wireless sensor network. *2008 International Conference on Intelligent Sensors, Sensor Networks and Information Processing*, pages 599–604, December 2008.

[46] G. Werner-Allen, K. Lorincz, M. Ruiz, O. Marcillo, J. Johnson, J. Lees, and M. Welsh. Deploying a wireless sensor network on an active volcano. *IEEE Internet Computing*, 10(2):18–25, March 2006.

[47] E. Woodrow and W. Heinzelman. SPIN-IT : A DATA CENTRIC ROUTING PROTOCOL FOR IMAGE RETRIEVAL IN WIRELESS NETWORKS. In *Proc. IEEE ICIP*. Citeseer, 2002.

[48] J. Wright, C. Gibson, F. Bergamaschi, K. Marcus, R. Pressley, G. Verma, and G. Whipps. A dynamic infrastructure for interconnecting disparate ISR/ISTAR assets (the ITA sensor fabric), 2009.

[49] Yong Yao and Johannes Gehrke. The cougar approach to in-network query processing in sensor networks. *ACM SIGMOD Record*, 31(3):9, September 2002.

[50] Yan Yu and Ramesh Govindan. Geographical and energy aware routing: A recursive data dissemination protocol for wireless sensor networks. *Technical Report, UCLA-CSD TR-01-0023*, 2001.

[51] Yang Yu, Bhaskar Krishnamachari, and VK Prasanna. Issues in designing middleware for wireless sensor networks. *Network, IEEE*, pages 1–7, 2004.

[52] Marco Zennaro and Antoine Bagula. Planning and deploying long distance wireless sensor networks: The integration of simulation and experimentation. *. . . and Wireless Networks*, pages 191–204, 2010.

[53] ZSL. Instant Wild Live Wildlife Photographs.