A Framework for Estimating the Energy Consumption Induced by a Distributed System’s Architectural Style

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ABSTRACT

The architectural style used in a software system is an important factor in satisfying its quality requirements. In battery-powered environments, such as mobile and pervasive systems, efficiency with respect to energy consumption is often an important quality requirement. In this paper, we present a framework that facilitates early estimation of the energy consumption induced by an architectural style in a distributed system, and enables an engineer to use energy consumption estimates along with other quality attributes in determining the most appropriate style for a given distributed application. We apply the framework to three architectural styles, and evaluate it for precision and accuracy using a middleware platform that supports the implementation of those styles. In a large number of application scenarios, our framework exhibited excellent precision, in that it was consistently able to correctly rank the styles and estimate the relative differences in their energy costs. Moreover, the framework has also proven to be accurate: its estimates were within 7% of each style implementation’s actual energy cost.

Categories and Subject Descriptors

D.2.11 [Software Engineering]: Software Architectures

General Terms

Design, Measurement

1. INTRODUCTION

A promising approach to addressing the challenges of developing distributed, mobile, and pervasive systems is to employ the principles of software architecture [9,12]. Software architecture provides abstractions for representing the structure, behavior, and key properties of a software system [20]. These abstractions include software components (computational elements), connectors (interaction elements), and configurations (specific assembles of components and connectors). Architectural styles (e.g., publish-subscribe, peer-to-peer, client-server) are key design idioms which further refine the vocabulary of components and connectors and propose constraints on how they may be integrated.

Architectural decisions made early in the design process are critical to successful development of a distributed system. In particular, selecting an appropriate architectural style has a significant impact on system quality attributes (e.g., latency, scalability, reliability, etc.). Energy efficiency is increasingly being defined as an important quality attribute for mobile and pervasive applications. However, there are currently no techniques for analyzing the impact of an architectural style on a system’s energy consumption. In fact, unlike other quality attributes, such as those mentioned above, a style’s energy consumption characteristics are not understood even in an informal and intuitive manner. In this paper, we address this problem via a framework that estimates the impact of a system’s architectural style on the system’s energy consumption. The framework is intended to be used during architectural design and enables an engineer to use energy consumption estimates, along with other quality attributes, in determining the most appropriate style for an application.

Figure 1 depicts the process and key artifacts employed by the framework. The framework defines a method to derive platform- and application-independent equations that characterize a style’s energy consumption behavior. We refer to the equations for a given style as that style’s energy cost model. Comparing the models of different styles yields insights into the essential differences between the energy costs induced by each style. We have derived the energy cost models for five architectural styles: client-server, peer-to-peer, C2, publish-subscribe (pub-sub), and pipe-and-filter[18]. For brevity, this paper shows the derivation of only the client-server and pub-sub styles.

Figure 1. High-level view of the energy consumption estimation framework.
The framework also defines a process for applying any style’s energy cost model to a given distributed system design, prior to system implementation. As we demonstrate, this is accomplished by gathering basic information about the target platform and the system design, and plugging these parameters into the energy cost model. We refer to an energy cost model that has been parameterized in this way as a energy prediction model. Energy prediction models enable scenario-specific comparisons of the energy costs of different styles, enabling architects to weigh trade-offs and determine the circumstances under which one style will be more efficient than another. We have created energy prediction models for four different distributed systems. For each system, we chose a set of candidate styles that were appropriate for the system, derived the energy prediction model for each candidate style, and evaluated the framework’s accuracy and precision under multiple execution scenarios.

In a recent short paper [17], we outlined the overview of our approach. This paper provides a detailed explanation of the framework and presents the evaluation results. In a large number of experiments, our framework exhibited excellent precision, in that it was correctly able to determine the scenarios under which one style is more efficient than another, and estimate the relative differences in their energy costs. The framework also proved to be accurate: it consistently produced energy consumption estimates that differed from the actual measured energy costs by at most 7%.

This paper is organized as follows: Section 2 describes how our framework defines energy cost models and derives the energy cost models for the client-server and pub-sub styles. Section 3 discusses how energy cost models can be applied to a distributed system, and applies the pub-sub energy cost model to an example application. We present evaluation results in Section 4. Section 5 describes related research, and Section 6 concludes the paper.

2. ENERGY COST MODELS

Fielding [2] and Mehta [13] identified more than twenty architectural styles for distributed systems. In this section, we first show a uniform way to derive a style’s energy cost model, which is a symbolic expression that represents the energy cost induced by using the style. Then, we illustrate the approach by deriving the energy cost models for the client-server and pub-sub styles, which embody a diverse set of distributed systems characteristics, such as distribution, concurrency, and so on. The derivation of other styles is given in [18].

2.1 Generic Energy Cost Model

As it is common in power modeling of operating systems [10, 23], our energy cost model consists of multiple linear equations. We model the energy consumed by a distributed system as the sum of the energy consumed by its constituent $n$ components and $m$ connectors, as shown in Equation 1.

$$\text{overallEC} = \left( \sum_{i=1}^{n} \text{EC}(\text{Comp}_i) \right) \ast \left( \sum_{j=1}^{m} \text{EC}(\text{Conn}_j) \right)$$

The energy cost of a component $\text{Comp}_i$ can be expressed as shown in Equation 2. In this equation, $E_{\text{logic},i}$ is the computational energy cost of the component $\text{Comp}_i$ due to executing its core business logic, while $E_{\text{commWithConn},i}$ represents the energy cost of exchanging data via connectors.

$$E(\text{Comp}_j) = E_{\text{logic},j} + E_{\text{commWithConn},j} \quad (2)$$

In this work, we assume that a component’s core business logic remains the same for all styles. We acknowledge that this logic may need to be refactored in some cases. For example, the logic required by a component to manage its interfaces might differ among styles. We account for these differences in $E_{\text{commWithConn},i}$ of Equation 2. This implies that the computational energy cost of a component (i.e., $E_{\text{logic},i}$ in Equation 2) remains the same across all candidate styles, so our framework does not require the actual value of $E_{\text{logic},i}$ to compare the energy consumption of multiple styles.

Similarly, the energy consumption of a connector $\text{Conn}_j$ can be expressed as in Equation 3.

$$E(\text{Conn}_j) = E_{\text{comm},j} + E_{\text{logic},j} \quad (3)$$

$E_{\text{comm},j}$ represents the energy consumption of communication, which includes the cost of exchanging data both locally or remotely. We can calculate $E_{\text{comm},j}$ as shown in Equation 4. $E_{\text{commWithComp},j}$ represents the energy consumed by exchanging data with components, while $E_{\text{remoteComm},j}$ and $E_{\text{localComm},j}$ are the energy costs of exchanging data with remote and local connectors, respectively.

$$E_{\text{comm},j} = E_{\text{commWithComp},j} + E_{\text{remoteComm},j} + E_{\text{localComm},j} \quad (4)$$

Equation 5 shows how to determine $E_{\text{logic},j}$, which is the energy cost of services (other than communication) that a connector may provide [14]:

- **Coordination** – Transfer of control among components.
- **Conversion** – Adaptation of interfaces or data provided by one component to that required by another.
- **Facilitation** – Mediation and streamlining of interaction.

$$E_{\text{logic},j} = E_{\text{coord},j} + E_{\text{conv},j} + E_{\text{facil},j} \quad (5)$$

By combining the above equations, we arrive at the generic energy cost model given in Equation 6. The parameters in Equation 6 vary according to style. In the following subsections, we illustrate how we can create style-specific energy cost models for the client-server and pub-sub styles by refining these parameters.

$$\text{overallEC} = \sum_{i=1}^{n} E_{\text{logic},i} + E_{\text{commWithConn},i} + \sum_{j=1}^{m} \left( E_{\text{commWithComp},j} + E_{\text{remoteComm},j} + E_{\text{localComm},j} \right)$$

$$+ \left( \sum_{j=1}^{m} E_{\text{logic},j} + E_{\text{comm},j} + E_{\text{facil},j} \right)$$

$$+ \left( \sum_{j=1}^{m} E_{\text{coord},j} + E_{\text{conv},j} + E_{\text{localComm},j} \right)$$
2.2 Client-Server Energy Cost Model

To refine the energy cost parameters introduced in our generic energy cost model shown in Equation 6, we characterize the communication, coordination, conversion, and facilitation required by the client-server style. Figure 2 shows an example of a distributed system designed in the client-server style. Connectors in this style are commonly implemented as middleware stubs and skeletons. The client-server style behaves as follows:

Component Communication: Clients send requests to and receive responses from client connectors. Servers exchange requests and responses with server connectors.

Connector Communication: Client connectors receive requests from clients and forward them to the appropriate server connector, and receive responses from server connectors and return them to clients. Server connectors receive and buffer requests from client connectors and forward them to servers, and receive responses from servers and return them to client connectors.

Connector Coordination: Connectors transfer execution control by passing requests and responses to clients and servers.

Connector Conversion: Connectors marshal/unmarshal requests and responses.

Connector Facilitation: Connectors create and manage connection objects that implement remote communication.

Equation 7 shows the energy cost $E_{\text{commWithConn},j}$ of a client $\text{Comp}_i$ due to sending requests to and receiving responses from a connector. $a_{ij}$ is the total number of requests made by the client. $E_{\text{toConn},k}$ and $E_{\text{fromConn},k}$ represent the energy costs due to sending the $k$th request to and receiving its response from the connector, respectively. $E_{\text{commWithConn},j}$ of a server $\text{Comp}_i$ can be calculated in the same manner using Equation 7. The values of $E_{\text{toConn}}$ and $E_{\text{fromConn}}$ depend on the platform-specific communication mechanism used between the relevant component and connector. For example, if a queue is used, the values of $E_{\text{toConn}}$ and $E_{\text{fromConn}}$ are constant; on the other hand, if IPC is used, they are directly proportional to the size of exchanged data [23]. We show how these values can be determined for a given platform in Section 3.

$$E_{\text{commWithConn},j} = \sum_{k=1}^{a_{ij}} \left( E_{\text{toConn},k} + E_{\text{fromConn},k} \right)$$  (7)

The energy cost $E_{\text{commWithComp},j}$ of a client connector $\text{Conn}_j$ incurred by receiving requests from and forwarding responses to clients can be calculated using Eq. 8. $b_{ij}$ is the total number of requests received from clients, while $E_{\text{fromComp},l}$ and $E_{\text{toComp},l}$ represent the energy costs due to receiving the $l$th request and sending the $l$th response. We can also calculate $E_{\text{commWithComp},j}$ for a server connector $\text{Conn}_j$ using Equation 8.

$$E_{\text{commWithComp},j} = \sum_{l=1}^{b_{ij}} \left( E_{\text{fromComp},l} + E_{\text{toComp},l} \right)$$  (8)

Data exchange between client and server connectors may be either local (if they reside on the same host) or remote, and we handle each case differently. Due to space constraints, we only show how to model remote communication here. The local case is given in [18]. To model the energy consumption due to remote exchange of data between client and server connectors, we assume that the total energy cost is proportional to the size of the exchanged data. This has been shown to be a highly accurate characterization for commonly used network protocols, such as UDP [1]. Based on this, $E_{\text{remoteComm},j}$ of a client connector $\text{Conn}_j$ due to sending $c_j$ requests and receiving responses can be estimated as shown in Eq. 9. $t\text{Size}_l$ and $r\text{Size}_l$ are the sizes (e.g., KB) of the $l$th transmitted request and its response. $tE$ and $rE$ are the energy costs (Joule/byte) of transmitting and receiving a unit of data, respectively. $tS$ and $rS$ represent constant energy overheads associated with channel acquisition [1]. All of these values can be easily measured for a given implementation platform, as we demonstrate in Section 3.

$$E_{\text{remoteComm},j} = \sum_{l=1}^{c_j} \left( (t\text{Size}_l \times tE) + (r\text{Size}_l \times rE + rS) \right)$$  (9)

Similarly, we can calculate $E_{\text{remoteComm},j}$ of server connector $\text{Conn}_j$ using Eq. 10. $d$ is the total number of requests received over the network, and $r\text{Size}_l$ and $t\text{Size}_l$ are the sizes of the $l$th received request and its responses, respectively. $E_{\text{buffer},l}$ is the energy cost of buffering the $l$th received request.

$$E_{\text{remoteConn},j} = \sum_{l=1}^{d} \left( (t\text{Size}_l \times tE + tS) + (r\text{Size}_l \times rE + rS) + E_{\text{buffer},l} \right)$$  (10)

In the client-server style, the cost $E_{\text{coord},j}$ due to coordination is not modeled separately because connectors transfer execution control by passing requests and responses, whose energy cost is already captured by $E_{\text{commWithConn},j}$.

The conversion cost $E_{\text{conv},j}$ of a client connector can be quantified as in Equation 11. $c_j$ is the number of requests sent remotely, while $E_{\text{mar}},$ and $E_{\text{unmar},l}$ are the energy costs of marshalling the $l$th request and unmarshalling its response, respectively. The conversion cost $E_{\text{conv},j}$ of a server connector can be calculated in the same manner.

$$E_{\text{conv},j} = \sum_{l=1}^{c_j} \left( E_{\text{mar},l} + E_{\text{unmar},l} \right)$$  (11)

The facilitation cost $E_{\text{facil},j}$ of client and server connectors is calculated with Equation 12. $E_{\text{remoteConn}}$ and $E_{\text{localConn}}$ are the energy costs due to establishing a single remote or local connection, respectively. $Num_{\text{remoteConn}},j$ and $Num_{\text{localConn}},j$ are the numbers of remote and local connections established, respectively.

$$E_{\text{facil},j} = \left( E_{\text{remoteConn}} \times Num_{\text{remoteConn}},j \right) + \left( E_{\text{localConn}} \times Num_{\text{localConn}},j \right)$$  (12)

By substituting Equations 7 - 12 into the generic energy cost model in Equation 6, we arrive at the client-server energy cost model.
cost model (omitted here due to space constraints). Note that
the energy cost parameters (e.g., $E_{\text{fromConn}}, E_{\text{fromComp}}, tEC$, etc.)
introduced in this section are platform-specific, i.e., their
values depend on the hardware, OS, and middleware on which
an application is deployed. We elaborate on how these
parameters are determined for an actual platform in Section 3.

2.3 Pub-Sub Energy Cost Model

The pub-sub style reduces coupling by providing
transparency and anonymity of component identities and
locations. Figure 3 shows an example of a distributed system
designed using the pub-sub style. A component in the pub-sub
style may be a publisher, a subscriber, or both. Pub-sub
constructs have the following characteristics:

**Component Communication:** Subscribers declare the
events they wish to receive by sending subscription requests to
a pub-sub connector. They then asynchronously receive the
published events of interest from the pub-sub connector. In
addition, subscribers can send unsupscription requests to stop
receiving certain events. On the other hand, publishers may
continuously and asynchronously publish events by
transmitting them to a pub-sub connector.

**Connector Coordination:** Pub-sub connectors provide
time decoupling among subscribers and publishers by
queueing events, so that they can be delivered later.

**Connector Conversion:** Pub-sub connectors marshal and
unmarshal events that are transmitted remotely.

**Connector Facilitation:** Pub-sub connectors (1) manage
subscriptions and publications, (2) find the set of subscriptions
that match each published event, and (3) create connection
objects that implement remote communication.

Based on the above characterization, we can first calculate
the energy cost $E_{\text{commWithConn},j}$ of a component $Comp_j$ due to
exchanging subscriptions, unsubscriptions, and events with
pub-sub connectors as follows:

$$E_{\text{commWithConn},j} = \sum_{k=1}^{a_j} E_{\text{fromConn},k} + \sum_{j=1}^{b_j} E_{\text{UnsubBuffer},n}$$

(13)

$a_j$ is the total number of events published by the
component, and $b_j$ represents the total number of events
received by the component. $E_{\text{fromConn},k}$ is the energy cost of
sending the $k$th event to a pub-sub connector, while $E_{\text{fromConn},l}$

is the energy cost of receiving the $l$th event from a connector.
$c_i$ and $d_i$ are the numbers of subscriptions and unsubscriptions
sent by the component. $E_{\text{subs},m}$ and $E_{\text{Unsub},n}$ are the energy
costs incurred by sending the $m$th subscription and $n$th
unsupscription to the pub-sub connector.

The energy cost $E_{\text{commWithConn},j}$ of a pub-sub connector
$Conn_j$, incurred by exchanging subscriptions and events with
components can be calculated as follows:

$$E_{\text{commWithConn},j} = \sum_{k=1}^{o_j} \left( (E_{\text{fromConn},k} + E_{\text{fromBuffer},k}) \times \right.$$  
$$\sum_{l=1}^{c_i} E_{\text{fromComp},l} + \sum_{m=1}^{d_i} E_{\text{UnsubBuffer},m} + \sum_{n=1}^{b_j} \left( E_{\text{UnsubBuffer},n} \times ight.$$  
$$\left. E_{\text{toConn},k} \right)$$

(14)

$$E_{\text{remoteComm},j} = \sum_{k=1}^{o_j} \left( \left( rSize_k + tEC + tS \right) \times \right.$$  
$$\sum_{l=1}^{c_i} E_{\text{fromComp},l} + \sum_{m=1}^{d_i} E_{\text{UnsubBuffer},m} + \sum_{n=1}^{b_j} \left( \left( rUnsubBuffer_m + tEC + tS \right) \times ight.$$  
$$\left. E_{\text{UnsubBuffer},m} \right)$$

(15)

$p_j$ is the total number of events received over the network,
$r_j$ is the total number of subscriptions received and $t_j$ is the
total number of unsubscriptions received. $q_j$ is the total number
of events sent over the network, whereas $s_j$ is the total number
of subscriptions and $u_j$ is the total number of unsubscriptions.
$rSize_e$, rSubSize_m and rUnsubSize_m are the sizes of the $k$th
event, $m$th subscription and $n$th unsubscription, respectively.
$E_{\text{buffer},k}$, $E_{\text{SubBuffer},m}$ and $E_{\text{UnsubBuffer},n}$ are the energy costs of
buffering the $k$th event, $m$th subscription and $n$th
unsupscription. Other parameters are the same in Eq. 9.
We can calculate $E_{localComm,j}$ of a pub-sub connector due to exchanging subscriptions, unsubscriptions, and events with local pub-sub connectors as follows:

$$E_{localComm,j} = \sum_{k=1}^{p_j} (E_{localReceiv,k} + E_{buffer,k}) + \sum_{l=1}^{q_j} E_{localTrans,l} + \sum_{m=1}^{r_j} (E_{localSubReceiv,m} + E_{buffer,m}) + \sum_{n=1}^{s_j} E_{localSubTrans,n} + \sum_{o=1}^{u_j} (E_{localUnsubReceiv,o} + E_{unsubBuffer,o}) + \sum_{p=1}^{z_j} E_{localUnsubTrans,p}$$

$p_j$, $q_j$, and $r_j$ are the total numbers of events, subscriptions and unsubscriptions received from local pub-sub connectors, respectively. $E_{localReceiv,k}$, $E_{localSubReceiv,m}$ and $E_{localUnsubReceiv,o}$ are the costs of receiving the $j$th event, $m$th subscription and $o$th unsubscription. $E_{buffer,k}$, $E_{buffer,m}$ and $E_{unsubBuffer,o}$ are the energy costs of buffering the $k$th event, $m$th subscription and $o$th unsubscription. $E_{localTrans,l}$ and $E_{localSubTrans,n}$ and $E_{localUnsubTrans,p}$ are the energy costs of sending the $l$th event, $n$th subscription and $p$th unsubscription to local pub-sub connectors, respectively.

The energy cost $E_{coordin,j}$ of a pub-sub connector incurred by performing coordination can be calculated as follows:

$$E_{coordin,j} = \sum_{k=1}^{x_j} E_{queue,k}$$

$x_j$ is the total number of events received by the connector, and $E_{queue,k}$ represents the energy cost due to queuing the $k$th event.

The conversion cost $E_{conver,j}$ can be calculated as follows:

$$E_{conver,j} = \sum_{k=1}^{y_j} E_{unmar,k} + \sum_{l=1}^{z_j} E_{unmarSub,l} + \sum_{m=1}^{t_j} E_{unmarUnsub,m} + \sum_{o=1}^{v_j} E_{unmarUnsub,o} + \sum_{p=1}^{w_j} E_{unmarUnsub,p}$$

$p_j$, $q_j$, and $r_j$ are the total numbers of events, subscriptions and unsubscriptions to be unmarshalled, whereas $s_j$, $t_j$ and $u_j$ are the total numbers of events, subscriptions and unsubscriptions to be marshalled, respectively. $E_{unmar,k}$, $E_{unmarSub,l}$ and $E_{unmarUnsub,m}$ are the energy costs of unmarshalling the $k$th event, $l$th subscription and $m$th unsubscription, respectively. $E_{unmarSub,o}$ and $E_{unmarUnsub,p}$ represent the energy costs of marshalling the $o$th event, $p$th subscription, and $q$th unsubscription, respectively.

Finally, the facilitation cost $E_{facili,j}$ of a pub-sub connector is:

$$E_{facili,j} = \sum_{k=1}^{x_j} E_{facili,k} + \sum_{m=1}^{y_j} E_{procSubs,m} + \sum_{o=1}^{z_j} E_{procUnsubs,o} = \left(\sum_{l=1}^{p_j} E_{RemoteConn^\times NumRemoteConn,l}\right) \times \left(\sum_{n=1}^{q_j} E_{LocalConn^\times NumLocalConn,n}\right)$$

Finally, the facilitation cost $E_{facili,j}$ of a pub-sub connector is:

$$E_{facili,j} = \sum_{k=1}^{x_j} E_{facili,k} + \sum_{m=1}^{y_j} E_{procSubs,m} + \sum_{o=1}^{z_j} E_{procUnsubs,o}$$

$x_j$ is the total number of events received by the connector, whereas $y_j$ and $z_j$ are the total numbers of subscriptions and unsubscriptions received by the connector, respectively. $E_{RemoteConn}$ is the energy cost of retrieving the set of subscribers for the $j$th event from a subscription database. $E_{procSubs,m}$ and $E_{procUnsubs,o}$ represent the energy consumption of processing the $m$th subscription and $o$th unsubscription, respectively.

As in the case of the client-server style, the additional energy cost parameters (e.g., $E_{subs}$, $E_{unsubs}$, $E_{procSubs}$, etc.) introduced in this section are platform-specific. We discuss how they are determined for an actual platform in Section 3.

### 2.4 Style-Induced Energy Trade-Offs

To compare the energy costs induced by architectural styles, and illuminate their fundamental differences, we derive the algebraic difference between their energy cost models. For example, by subtracting the equation representing the total energy cost of the pub-sub style from that of the client-server style, a number of terms cancel out, leaving terms representing the costs of connector communication and facilitation. Inspection of these terms clearly reveals that the number of messages exchanged in the two styles is different, and the pub-sub style inures a facilitation cost that is not present in the client-server style. The intuition behind this result is that the client-server style is based on a point-to-point interaction between components, while the pub-sub style may route messages more efficiently. Therefore, if a message is intended to be received from a remote host by multiple components running on one host, the client-server style in principle requires each message to be transmitted to each recipient component separately. On the other hand, the same could be achieved in the pub-sub style with only one transmission of the message. In this case, it is reasonable to expect the remote communication energy cost of client and server connectors (i.e., $E_{remoteConn}$ of Equations 9 and 10) to be larger than that of pub-sub connectors (i.e., $E_{remoteConn}$ of Equation 15). However, in general, pub-sub connectors incur a higher facilitation energy cost (i.e., $E_{facili}$ of Equation 19) than client and server connectors (i.e., $E_{facili}$ of Equation 12) because the pub-sub style has the additional overhead of managing subscriptions and retrieving them for each published event.

Therefore, through the systematic determination of energy cost parameters, as outlined above, it becomes immediately clear that the energy trade-off between the client-server and pub-sub style is dominated by two factors: (1) the number of separate messages exchanged remotely and (2) the facilitation overhead of a pub-sub connector. In other words, a client-server application consumes more energy than the same application implemented in the pub-sub style only if the energy cost of exchanging additional messages is larger than the facilitation cost of the connectors in pub-sub, and vice versa.

This type of comparison can be conducted for arbitrary architectural styles whose energy costs have been modeled using the process described in this paper. An important contribution of our framework is that it allows architects to intuitively understand the energy trade-offs between different styles based solely on the style-induced characteristics, and irrespective of the implementation platform.

### 3. ENERGY PREDICTION MODELS

The energy cost models given in Section 2 provide a symbolic representation for the energy cost induced by the architectural style of a distributed system. However, the actual energy consumption induced by various styles is dependent on several platform-specific and application-specific properties. In this section, we show how to apply the energy cost model...
for an architectural style to a distributed system by creating a specific energy prediction model. To derive an energy prediction model, the following information is necessary:

1. **Platform-specific energy cost model parameters** must be determined through mapping the energy cost model parameters to the target implementation platform and measuring their actual costs. These parameters include the cost of transmitting data over a network and the cost of performing lookups in routing tables. The process for performing this task is explained in Section 3.1.

2. **Application-specific energy cost model parameters** must be determined from the system design. These parameters include the system’s components, connectors, their configuration, sizes of exchanged messages, etc. The process for gathering these parameters is detailed in Section 3.2.

Note that determining the values of the above sets of parameters only requires access to the target platform and basic application design information, and does not require the actual implementation of the application. This allows our framework to be used by architects early in the design process.

### 3.1 Platform-Specific Model Parameters

To determine the energy costs represented by platform-specific parameters in an energy cost model, we characterize these costs in terms of interfaces provided by the underlying platform. This process consists of two steps: (1) mapping each platform-specific parameter to interface(s) of the underlying platform that incur that cost, and (2) measuring with a digital multimeter the actual energy consumed when those interfaces are invoked. We have mapped energy cost models to three different platforms: Prism-MW [12], Java RMI [6] and TAO [16]. Due to space constraints, here we only describe the mapping for the pub-sub energy cost model onto Prism-MW. The process for other styles and platforms is analogous, and the details of these other mappings are given in [18].

Prism-MW, a lightweight, component-based middleware platform, is an appropriate demonstration and evaluation platform for our framework for two reasons: (1) it is intended for resource-constrained and mobile systems, to which our work is directly relevant; and (2) it supports for numerous architectural styles [12], giving us a common platform to evaluate the framework’s utility. Prism-MW provides programming language-level constructs that directly correspond to and implement software architecture-level concepts such as components, connectors, topologies, and ports. In general, there are many ways of implementing an interface; in Prism-MW, the invocation of an interface corresponds to an event being passed to the handle method of one of the Prism-MW constructs. Therefore, we map the energy cost model parameters for the pub-sub style to the Prism-MW platform in the following way:

- The parameters $E_{subs}$, $E_{unsubs}$, $E_{procSubs}$ and $E_{procUnsubs}$ (of Eq. 13 and 19) are incurred when a pub-sub connector handles a subscribe or unsubscribe event.
- The parameters $E_{route}$, $E_{isObject}$, $E_{fromComp}$ and $E_{recBuffer}$ (of Eq. 13, 14, 15 and 19) are incurred when a pub-sub connector handles a published event.
- The parameters $E_{fromConn}$ and $E_{toComp}$ (of Eq. 14) are incurred when a component handles an event.

Once cost model parameters have been mapped to interfaces of the target platform, it is necessary to measure the actual energy cost of invoking each interface, using the measurement setup shown in Figure 4. A benchmarking application invokes each interface with different input parameter values and measures the current drawn from a power supply. We then apply multiple regression to the values recorded by the benchmark application to estimate the relationship between the input variables and the energy consumed. This relationship depends on the target platform. As one target platform, we used a Compaq iPAQ 3800 mobile device running embedded Linux with a 206MHz Intel StrongARM processor, 64MB memory, and 11Mbps 802.11b compatible wireless PCMCIA card. We chose a version of Prism-MW that runs on top of the JamVM 1.4.5 [7], which is a lightweight Java Virtual Machine. The iPAQ device was connected to a 5V DC power supply and HP 3458 digital multimeter. The multimeter sampled the current drawn by the iPAQ at a high frequency. A data collection computer controlled the multimeter and recorded the current samples. Table 1 summarizes the measured energy consumption values.

Once platform-specific parameter values have been obtained, the remaining undefined values in the energy cost model are application-specific parameters. The next section describes how to extract this information from a system design to create a completely parameterized energy prediction model.

#### 3.2 Application-Specific Model Parameters

Energy cost models contain parameters related to the number and size of messages exchanged between components. Also, whether messages are transmitted locally or over a network has a major impact on energy use. Determining these application-specific parameters requires extracting the following information from the system design:
• The system components, connectors, and configuration
• The deployment architecture, which is the allocation of the components and connectors to hardware hosts
• Sizes of messages exchanged between components
• A set of destination components for each message

To illustrate how to derive an energy prediction model, given the above information, consider the sensor application shown in Figure 5. The application was designed using two architectural styles: client-server and pub-sub. The FireAlarmReceiver component on Gateway 1 translates and aggregates alarms received from fire detection sensors periodically, and propagates them to the FireAlarmAnalyzer on the Hub. Additionally, these alarms are logged by the AlarmLogger. The FireAlarmAnalyzer interprets the alarm data to determine whether there is actually a fire. If the FireAlarmAnalyzer concludes that there is a fire, it transmits a sensor-activation message to the FireAlarmReceiver, which in turn sends an activation signal to all fire sensors. An analogous processing path takes place in the intrusion detection sensors, IntrusionAlarmReceiver, and IntrusionAlarmAnalyzer.

In the client-server architecture, the Receiver components act as clients and invoke interfaces on the Analyzer and Logger via their local connectors. The client connectors on Gateways 1 and 2 transmit requests received from local Receiver components to the Analyzer and Logger separately, which indicates that each alarm requires two remote transmissions. On the other hand, in the pub-sub architecture, the Analyzer components subscribe to fire or intrusion alarm events, and the AlarmLogger subscribes to both event types. When the FireAlarmReceiver publishes a fire alarm event, the pub-sub connector retrieves the event’s subscribers (i.e., FireAlarmAnalyzer and AlarmLogger) and routes the event to them, which requires only one transmission from Gateway 1.

Thus, to create an energy prediction model for the pub-sub instance of the above application, an architect characterizes the system in the following way:
• 5 components and 3 pub-sub connectors run on 3 hosts.
• The types of messages are fire alarm, intrusion alarm, fire sensor activation, intrusion sensor activation, and subscription request.
• The destination components for a fire alarm message are FireAlarmAnalyzer and AlarmLogger, while an intrusion alarm message has the destinations of IntrusionAlarmAnalyzer and AlarmLogger. The destination of a fire sensor activation message is the FireAlarmReceiver, whereas an intrusion sensor activation has the IntrusionAlarmReceiver destination.

3.3 Style-Induced Energy Trade-Offs

Instantiating a style’s energy cost model with platform-specific and application-specific energy cost model parameters leads to an energy prediction model that can be used to estimate the relative differences between styles with respect to energy consumption. For example, in the sensor application scenario, the energy cost incurred on the Gateway by the occurrence of a fire alarm event (in mJ) is given in Equations 20 (for the pub-sub style) and 21 (for the client-server style).

\[ 87.85 + 2.752 \times \text{eventSize} \]  \hspace{1cm} (20)

\[ 43.39 + 5.504 \times \text{requestSize} + 38.07 \times \text{replySize} \]  \hspace{1cm} (21)

Thus, given rough estimates of the sizes of the various events in the scenario, an architect can determine which style is more efficient. For example, if we assume the sizes of all application events to be 9 KB, the cost of each fire alarm is 112.62 mJ for pub-sub and 177.15 mJ for client-server. If we assume subscription and connection setup (i.e., facilitation) to be one-time costs, over time the pub-sub architecture would be approximately 37% more efficient. More generally, an architect can easily vary any of the parameters in the model — hosts, components, messages sizes, etc. — and investigate which style is most efficient in different situations.

4. EVALUATION

This section presents our framework evaluation. To check the accuracy of our framework’s estimates, we measured the actual energy costs induced by the architectural styles for our application scenarios and compared these values to energy consumption predictions made by our framework.

We evaluated our framework using four applications: the sensor application described in Section 3.2, a mobile sales application, a search-and-rescue coordination application, and an XML data streaming application. For each application, we used our framework to estimate the energy consumed on each host and calculated the overall energy consumption induced by each style by summing up the hosts’ energy costs. We then measured the actual amount of energy consumed by the implemented system, again using the digital multimeter setup described in Section 3.2. We varied the frequencies and sizes of data exchanges and component interactions stochastically.

For the sensor application, the framework suggested that the pub-sub style would be significantly more energy efficient than the client-server style. As shown in Figure 6, the energy consumption estimates of the framework fell within 7% of the measured energy costs for both styles. Moreover, as our framework predicted, the pub-sub style was much more energy-efficient for this scenario because the energy cost incurred by exchanging more data remotely in the client-server style exceeds the energy overhead due to processing the subscriptions and publications in the pub-sub style.

As shown in Figure 6, the energy consumption predictions for the mobile sales application made by our framework were also within 7% of the actual measured costs for both styles. The framework also correctly predicted that the client-server
style is more energy-efficient than the pub-sub style. The same level of accuracy was observed on the other two applications. The detailed data for all applications is given in [18].

6. CONCLUSION
We presented a framework that facilitates early estimation of the energy consumption induced by an architectural style on a distributed software system. This capability enables an engineer to employ energy cost predictions along with other quality attributes in determining the most appropriate architectural style for a given distributed application before the implementation of the system. Our extensive evaluation of the framework with respect to accuracy in a large number of distributed application scenarios has shown an error bound of at most 7% as compared to the actual energy cost.

7. REFERENCES