

Safe Upper-bounds Inference of Energy Consumption for Java Bytecode Applications

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Motivation

- Many space applications handle a large amount of data and its analysis is often critical for the underlying scientific mission
- Transmitting data to the remote control station is usually too expensive
- Instead, modern space applications are increasingly relying on autonomous on-board data analysis
- Examples of these applications can be: sensor networks, on-board satellite-based platforms, on-board vehicle monitoring system, etc.
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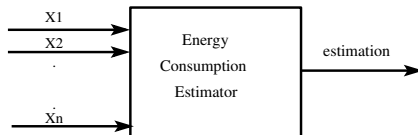
A key requirement is to **minimize energy consumption**

Related Work

In current systems the estimation of energy consumption is inferred at *run-time* generating often large sets of random inputs $S = \{X_1, \dots, X_n\}$ for calculating the energy consumption.

Φ : true energy consumption

$\hat{\Phi}(S)$: estimated (from S) energy consumption

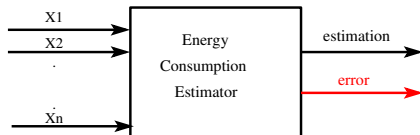


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Disadvantages:

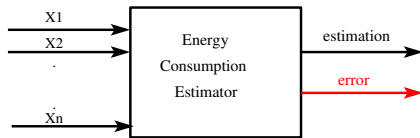
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- There is always a potential **error** ($\hat{\Phi}(S) - \Phi$)

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But safety critical systems require more **formal techniques** for inferring a safe energy consumption estimation.

Our Approach

We propose a fully automated analysis that infers **safe upper bounds** on the energy consumption in terms of input data sizes for Java (bytecode) applications **at compile time**

- 1 Define *energy consumption model* \mathcal{M} [LL07] that describes the *upper bound cost* of each bytecode inst. in terms of joules it consumes:

Opcode	Inst. Cost in μJ	Mem. Cost in μJ	Total Cost in in μJ
iadd	.957860	2.273580	3.23144
isub	.957360	2.273580	3.230.94
...

- 2 With this resource model, we then generate energy consumption cost equations using resource usage analysis [NMLH08] which are then solved returning safe, upper bound *energy cost functions*.

Generate *cost equation system* by abstracting the iterative constructs (loops and recursion), and by inferring *size relations* between the arguments.

```
public int fact(int n) {  
    if (n == 0) {  
        return 1;  
    } else {  
        return n * fact(n - 1);  
    }  
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$$\mathcal{S}_{ret}(s_n) = s_n \times \mathcal{S}_{ret}(s_n - 1)$$

Equation systems can be often solved by usually using *difference equation solvers*, thus obtaining a *closed form* solution.

$$\begin{aligned} \mathcal{S}_{ret}(0) &= 1 \\ \mathcal{S}_{ret}(s_n) &= s_n \times \mathcal{S}_{ret}(s_n - 1) \Rightarrow \mathcal{S}_{ret}(s_n) = s_n! \end{aligned}$$

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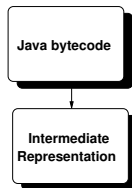
$$E_{fact}(0) = \underbrace{\mathcal{M}_{n=0}}_a(\mu J)$$

$$\underbrace{\mathcal{M}_{n>0}}_b(\mu J) + E_{fact}(s_n - 1)$$

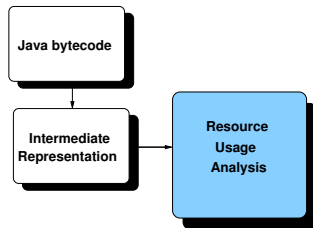
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$$\begin{aligned} E_{fact}(0) &= a \\ E_{fact}(s_n) &= b + E_{fact}(s_n - \vec{1}) \end{aligned} \quad E_{fact}(s_n) = a + (b \times s_n) (\mu J)$$

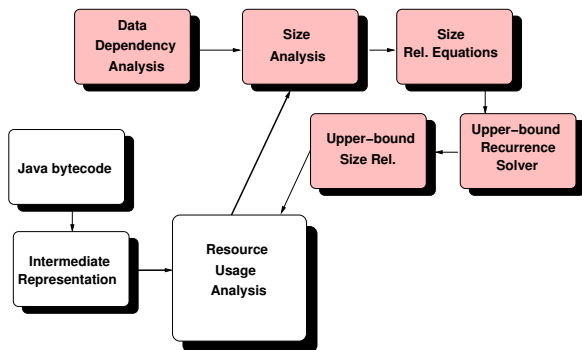
Energy Consumption Framework



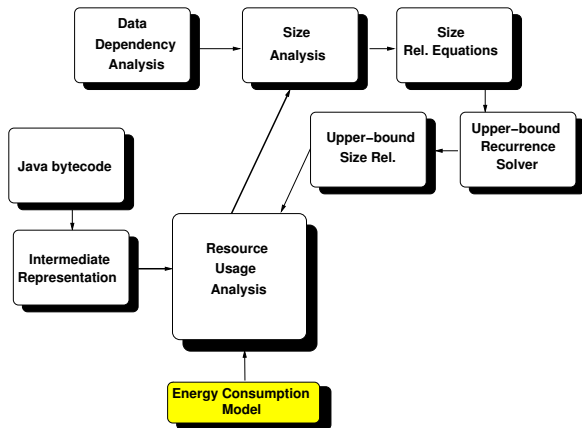
Energy Consumption Framework



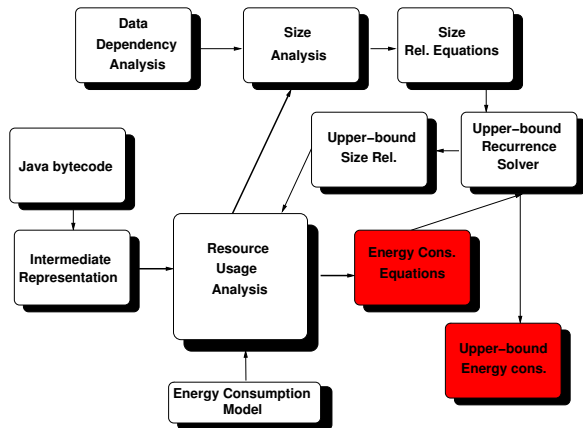
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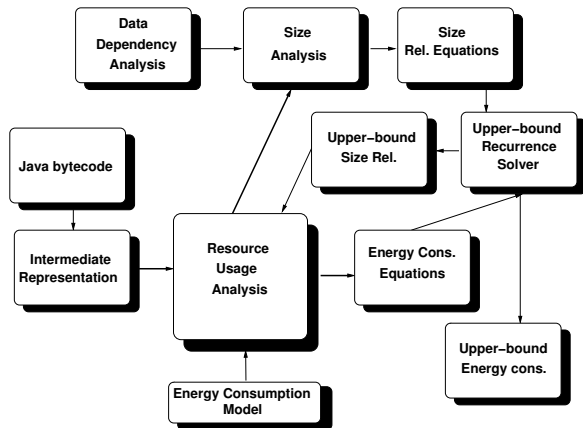
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Example: Java Program

```
import java.lang.Stream;
public class SensorNet {
    public StringBuffer collectData(Sensor sensors []) {
        int i;
        int n= sensors.length;
        StringBuffer buf = new StringBuffer();
        for (i=n; i > 0;i--){
            String data = sensors[i].read();
            buf.append(data);
        }
        return buf; }
    interface Sensor { String read();}
    class TempSensor implements Sensor {
        @Cost("10*size(ret)")
        public native String read(); }
    class SeismicSensor implements Sensor {
        @Cost("20*size(ret)")
        public native String read(); }
}
```

Example: Upper-bound Inference of Energy Consumption

- 1 Inference of size relationship equations:

$$S_{ret}(S_{this}, S_i, S_{sensors}) \leq \begin{cases} 1 & \text{if } s_i = 0 \\ S_{data} + S_{ret}(S_{this}, S_i - 1, S_{sensors}) & \text{if } s_i > 0 \end{cases}$$

- 2 Solving size relationship equations:

$$S_{ret}(S_{this}, S_i, S_{sensors}) \leq S_{data} \times S_i$$

- 3 Inference of energy consumption equations:

$$E_{collectData}(S_{this}, S_i, S_{sensors}) \leq \begin{cases} 241 & \text{if } s_i = 0 \\ 20 \times S_{data} + 487 + E_{collectData}(S_{this}, S_i - 1, S_{sensors}) & \text{if } s_i > 0 \end{cases}$$

- 4 Solving energy consumption equations:

$$E_{collectData}(S_{this}, S_i, S_{sensors}) \leq (20 \times S_{data} \times S_i) + (487 \times S_i) + 241$$

Conclusions

- We have defined and implemented an energy consumption analysis that:
 - ▶ infers relatively accurate safe upper bounds
 - ▶ is independent from the Energy Consumption Model \mathcal{M}
 - ▶ supports a reasonable set of data-structures (trees, arrays, lists, etc.) and standard Java libraries used in real applications
 - ▶ covers a good range of complexity functions ($O(1)$, $O(\log(n))$, $O(n)$, $O(n^2)$. . . , $O(2^n)$, . . .) and different types of structural recursion such as simple, indirect, and mutual.
- Many potential improvements (e.g., supporting more complex data-structures, more sophisticated data size metrics, etc.).

Questions ?

Bibliography



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