A Model-Based Approach to Designing QoS Adaptive Applications

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Motivation and Background

- Applications are distributed and network centric
- Stringent QoS requirements, including predictable and efficient data transfer and control
- Resources are constrained and shared
- Operate in dynamic environments

An emerging trend of customized critical distributed real-time embedded systems that are required to operate in highly dynamic networked environments

The need to handle dynamic conditions effectively with predictable quality of service

Gracefully handle degraded and hostile situations

Effectively utilize additional resources

Adapted from “The Future of AWACS”, by LtCol Joe Chapa
Our Goals

1. Identify essential elements of QoS adaptation that need to be captured in the modeling tool

2. Insert notations of these elements in design tools to enable quality-of-service modeling

3. Simplify the development of end-to-end and system-wide QoS control and adaptation for DRE systems

4. Improve the formalization, usability, and robustness of QoS adaptive concepts

5. Provide automated code synthesis and model refinement capabilities

6. Develop tools for predictable, controlled behavior and graceful degradation and recovery
Technology Bases – QuO Adaptive Middleware and GME Metamodelling Environment

- **QuO** is a middleware framework for developing QoS adaptive distributed systems
  - Works with CORBA, Java RMI
  - Provides measurement, control, and contract-driven adaptation
  - Provides a QoS behavior encapsulation model, called *qoskets*
  - Available open source at http://quo.bbn.com

- **GME** is a framework for creating domain-specific design environments
  - Supports abstract modeling concepts generic enough for a wide range of domains
  - Uses metamodeling to define the family of domain models that can be created
  - Supports tool integration and extensibility
  - Available at http://www.isis.vanderbilt.edu/Projects/gme/default.html
• Develop a modeling language supporting the high-level design of adaptation strategies in GME
• Map high-level specifications to run-time interfaces and constructs, such as QuO
• Feedback loop for incorporating runtime information into refinement of design-time model
• Simulation and analysis of adaptation strategies
• Demonstrate and evaluate in different DRE applications
Dissecting the Constituent Elements of QoS Adaptation

QoS aspects define a multi-dimensional adaptation space.

QoS knobs determine what can be changed.

QoS dimensions are not independent, knobs frequently affect more than one, e.g.,
- compressing data decreases network bandwidth but increases CPU usage
- increasing security commonly affects performance

The domain and mission define an upper bound on the QoS, e.g.,
- full motion video
- highest resolution a sensor can provide

and a lower bound, e.g.,
- minimum useful resolution
- lowest useful rate

The goal is to support the design of QoS adaptation separate from the functionality and at a higher level than programming
- Simplify the design of QoS adaptation
- Enable a larger class of practitioners

Model and synthesize adaptive behaviors that
- Ensure predictable, controlled behavior
- Satisfy mission requirements (i.e., use requirements to choose suitable tradeoffs)
- Graceful degradation and recovery
Essential Elements in QoS Adaptation

- **Mission requirements** - What does the system need to achieve? This is used to capture the high-level mission requirements of the system.

- **Observable parameters** - Where is the system within the QoS space? Before any adaptation action can be taken, the system has to know its current state with regard to the QoS of interest.

- **Controllable parameters and adaptation behaviors** - How can the system move through the QoS space? These are the knobs available to the application for QoS control and adaptation.

- **System dynamics** - Where can the system move within the QoS space? Based on the current state of the system (observable parameters) and the set of knobs available (controllable parameters and adaptation behaviors), there could be several options for adaptation.

- **Adaptation strategy** - Where should the system go within the QoS space? Here we actually choose an adaptation strategy, which specifies the adaptations employed and the tradeoffs made in response to dynamic system conditions in order to maintain an acceptable mission posture.
Distributed QoS Modeling Environment (DQME)

- Prototype DQME in the GME modeling environment captures all QoS essential elements identified:
  - Mission -> Mission requirements
  - QoSParam -> Observable and Controllable QoS parameters
  - SystemDynamics -> System dynamics
  - Controller->Adaptation strategies
- Works with a modeling language supporting Application Structure (SRML)
- Code synthesis through model interpreters
Application of DQME to Optimization of Signal Analyzers

• Our focus has been on the design stage optimization (vs. deployment stage) of signal analyzers, using adaptation and control to measure, predict, and tune parameter values for signals with known ground truth
  – Experiment 1 – Optimized parameters for PSK signal recognition (PSK, non-PSK)
  – Experiment 2 – Optimized parameters for PSK and FSK signal recognition (PSK, FSK, other)
  – Experiment 3 – Build an optimized signal analyzer for PSK, FSK, and multi-carrier PSK

• Signal Analyzer Architecture
  – Build on feature extractors - yield feature values that are useful for discriminating between different signal types and identifying signals of interest
  – OpBlocks - low-level signal processing operations in feature extractors
  – A Classifier - uses the feature values to identify the class of the incoming signal
  – OpBlocks and the Classifier itself can be parameterized and their values can be adjusted to improve the classification of the signal analyzer.
Signal Analyzer Model in DQME

Search space “controller” – tweaks parameter values systematically, checks against ground truth, and computes utility

Parameters and their references into OpBlocks

“Defender” View (thanks GME 4!)

Feature Extractor
Structure of Feature Extractor – constructed of OpBlocks
Structure of Signal Analyzer

Feature Extractor Model in DQME
QoS Modeling

Classification Outcome Definitions

<table>
<thead>
<tr>
<th>True Signal Type</th>
<th>Classification Output</th>
<th>Cost Factor for Classification Outcomes</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>PSK</td>
<td>FSK</td>
</tr>
<tr>
<td>PSK</td>
<td>Hit</td>
<td>FalseAlarm</td>
</tr>
<tr>
<td>FSK</td>
<td>FalseAlarm</td>
<td>Hit</td>
</tr>
<tr>
<td>Other</td>
<td>FalseAlarm</td>
<td>FalseAlarm</td>
</tr>
</tbody>
</table>

Therefore, the QoS can be defined as:

\[ QoS = 1 - \frac{1}{N} \sum_{i=1}^{N} v_i \]

where \( i \) is the index of the current signal, \( N \) is the size of the window, and \( v_i \) is the cost associated with the processing of the \( i \)th signal. Our adaptation strategy is then to maximize QoS for each signal set and minimize the cost associated with each signal classification by adjusting tunable (Controllable) parameter values.
## Signal Analyzer Optimization Results

<table>
<thead>
<tr>
<th>Experiment</th>
<th>Problem</th>
<th>Measure of QoS</th>
<th>Experiment Parameters</th>
<th>What Experiment Evaluated</th>
<th>Metrics (Qualitative/Quantitative)</th>
<th>Significant Results</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Recognize PSK signals</td>
<td>Percent of correct recognition</td>
<td>non-optimized algorithms and initial parameter values</td>
<td>Improvement in QoS after parameter tuning using local search</td>
<td>3 parameters tuned Total 540 searches Initial QoS: .750 Final QoS: .971</td>
<td>10x reduction in error (incorrect classification reduced from .25 to 0.029)</td>
</tr>
<tr>
<td>2</td>
<td>Recognize PSK and FSK signals</td>
<td>Percent of correct recognition</td>
<td>non-optimized algorithms and initial parameter values</td>
<td>Improvement in QoS after tuning; speedup of global search + grouping; tuning of more params</td>
<td>Initial QoS: .6875 4 params (ls): 1620 searches, final QoS: .8819 4 params (gs): 93 searches, final QoS: .845 10 params (gs): 429 searches, final QoS: 0.8608</td>
<td>2.2x reduction in error (error reduced from .3125 to .1392) 2.5x parameters in ½ the time</td>
</tr>
<tr>
<td>3</td>
<td>Identify PSK, FSK, multi-carrier PSK</td>
<td>Percent of correct recognition</td>
<td>system parameters with near optimal values</td>
<td>Reduction in manual tuning effort Improvement in QoS Ease of use</td>
<td>3 parameters tuned 9 hours labor to tune by hand; 2.25 labor to tune using tool Initial QoS: 89.12% Final QoS: 89.22% Some usage issues identified</td>
<td>4x reduction in manual labor 1% improvement over optimized values</td>
</tr>
</tbody>
</table>
Application of DQME to the PCES Capstone Demonstration

End-to-End Mission-Driven QoS Management

Surveillance
- Maximize surveillance area
- Sufficient resolution in delivered imagery to determine items of interest

Target Tracking
- UAV observing target provides high resolution imagery so that target or threat identification is possible

Battle Damage Assessment
- UCAV must provide high resolution imagery until a human operator has determined that it is sufficient
- UAV over target area must continue to provide target acquisition and engagement mission

The challenge is to program the dynamic control and adaptation to manage and enforce end-to-end QoS.

Role-defined requirements and tradeoffs (e.g., rate, image size, fidelity)
- Heterogeneous, shared, and constrained resources

Changing modes, participants, roles, and environmental conditions

Tactical links are statically allocated and unmanaged

Avionics platforms are closed with statically scheduled tasks

C2 and CAOC CPUs have variable dynamic load

LAN/WAN links are shared with (a priori) unknown load
Component Structure of the PCES Capstone Demonstration DQME Model

- Five RUAVs
- One UCAV
- CAOC
- Fire Effects Cell (Army Command)
- Local resource managers (LRMs)
- A System Resource Manager (SRM)
Mission Requirements of the PCES Capstone Demonstration

• Three participant roles:
  – **Surveillance**
  – **Target Tracking** – Observing an area of interest
  – **Battle damage assessment (BDA)**

• Each role specifies minimums and maximums
  – Minimum rate in **surveillance** must ensure no gaps in surveillance (function of the UAV camera scan size, altitude and speed of the UAV)
  – Minimum size (360x240) and resolution determined by domain experts

• Tradeoffs within roles
  – **Surveillance** adjusts rate of imagery, resolution (i.e., compression), followed by size
  – **Target Tracking** adjusts rate and allows cropping (if centered on area of interest)
  – **BDA** adjusts rate, size, and allows tiling of images
Observable and Controllable Parameters and Adaptation Behaviors

We have a set of off the shelf encapsulated behaviors (*qoskets*)

- CPU Broker (Univ. Utah)
- Compression
  - Wavelet, JPEG, PNG
- Cropping
- Setting DiffServ Codepoints
- Pacing (changing frame rate)
- Scaling
- Tiling
Multi-Layered Control

• System Resource Manager
  – A supervisory controller that responds to changes in role and relative priority within a role
  – Allocates available resources to participants
  – Provides policy of role, relative priority, bandwidth allocation, and CPU allocation to participants (UAVs, UCAV)

• Local Resource Managers
  – Determine how to effectively use the resources allocated to a participant to perform its role
  – Select an appropriate CPU reservation, network priority, and data manipulation (rate, compression, size) to fit the allocated resources and fulfill its role
  – Monitors actual resource usage and adjusts parameters to keep usage within allocation
  – Bases decisions on allocated resources, tradeoffs (from mission requirements), available adaptation strategies, and system dynamics

System Dynamics
compute resource unit

$$\begin{align*}
  \text{bw}_s\_\text{low} &= \text{num}_s\*\text{bw}_s/(\text{num}_s\_\text{urgent}\*\text{urgent}\_\text{wt} + \text{num}_s\_\text{high}\*\text{high}\_\text{wt} + \text{num}_s\_\text{normal}\*\text{normal}\_\text{wt} + \text{num}_s\_\text{low}\*\text{low}\_\text{wt})\*\text{low}\_\text{wt}; \\
  \text{bw}_s\_\text{normal} &= \text{bw}_s\_\text{low}\*\text{normal}\_\text{wt}/\text{low}\_\text{wt}; \\
  \text{bw}_s\_\text{high} &= \text{bw}_s\_\text{low}\*\text{high}\_\text{wt}/\text{low}\_\text{wt}; \\
  \text{bw}_s\_\text{urgent} &= \text{bw}_s\_\text{low}\*\text{urgent}\_\text{wt}/\text{low}\_\text{wt};
\end{align*}$$
PCES Capstone Demonstration Code Generation

- System-level resource allocation strategies are captured in SRM
- Model interpreter scans and extracts information from the model
- Generate equivalent C++ code that implements the logic in the model
- Tested and successfully used in SRM CIAO component in the capstone demo application
- We are working on generating more code, including the LRM

```
void SystemResourceManagement::AllocateResource(
    PCES_UAV::Priority priorityValue,
    PCES_UAV::Role roleValue,
    double available_bw,
    double available_cpu,
    double& allocated_bw,
    double& allocated_cpu,
    long& code_point)
ACE_THROW_SPEC(( CORBA::SystemException )){
    AllocateDSCP(code_point,roleValue);
    getRoleBW(available_bw);
    getRoleCPU(available_cpu);
    updateBDABW();
    updateSurveillanceBW();
    updateTargetTrackingBW();
    updateBDACPU();
    updateSurveillanceCPU();
    updateTargetTrackingCPU();
    AllocateBW(allocated_bw,roleValue,priorityValue);
    AllocateCPU(allocated_cpu,roleValue,priorityValue);
}
```

Generated SRM code

Logic in SRM model
Conclusions

• We have designed, built and evaluated significant elements of a QoS adaptive modeling environment
  – Developed domain specific, semantically rich modeling language (DQME) that allows QoS adaptation modeling in distributed, real-time, embedded systems
    » Identified essential elements of dynamic QoS adaptation
    » Demonstrated the ability to model QoS in different application domains
  – Code generation capability through GME interpreters and code-level QoS adaptation support through QuO middleware
    » Simulation code (MATLAB) for signal analysis
    » Runtime code (C++, CORBA, QuO) for both signal analysis and PCES capstone demo
  – Works with other tools in a combined tool chain, including SRML and CADML
• Generally made QoS adaptation design easier and provide productivity/quality improvement
• Remaining difficult problems
  – User interface issues, making modeling in DQME easier
  – Extraction of patterns of use