

Automation of Large-scale Computer Cluster Monitoring Information Analysis

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Motivation



- Demand on computational and storage resources is increasing
- Hardware limitations push to deploy very complex, heterogeneous, and massive computing and storage facilities
- Multiple software systems run together to provide data management and computing services
- The workload and environment conditions tend to change very rapidly with time - hard to detect the complex service status or foresee its reliability
- Hundreds of tuning parameters for each system limitations in experienced man power





- In time status identification
- Proper root cause analysis of the failure/problem
- Failure/degradation forecasting

Autonomic Computing Concept



Goal

Design an intelligent system to increase reliability, autonomy and performance of the computing facility (term arose in 2001)

- Self-configuration: Automatic configuration of components
- Self-optimization: Automatic discovery, and correction of faults
- Self-healing: Automatic monitoring and control of resources to ensure the optimal functioning with respect to the defined requirements
- Self-protection: Proactive identification and protection from arbitrary attacks

Autonomic Computing Structure



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We need to transform monitoring data to information about services

- Monitoring most important for all components
- Analyze to transform monitoring data in a knowledge about the system

Core Technique for Data Analysis



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Adaptive Network Based Fuzzy Inference System - ANFIS

- Widely used for nonlinear system identification
- Used for chaotic and sequential time series prediction
- Efficient to analyze qualitative and quantitative aspects of data



Linear Support Vector Machine (SVM) for feature selection/ranking

- Well known and adapted method in Genetics, Climatology, Finances
- In a number of studies SVM show superior results



Service Status Identification



Conditions

- Monitoring data sources: Ganglia monitoring tool and dCache log file
- Service of interest: The PnfsManager the core service of the dCache storage system
- Data collection period: 17/07/2013 08/08/2013, 18 failures registered
- Obstacle: Small amount of data for ANFIS training required feature selection



Feature selection





Feature Selection by SVM (using Java based software Rapidminer)



PnfsManager Status Identification





Metric	Values	Details
Precision	99%	Positive Predicted Values
Recall	99%	True-Positive rate
Specificity	88%	True-Negative rate
Accuracy	99%	-

Key metric for service functionality



Service Response Time - SRT

SRT reflects information from internal and external performance metrics for a service

Availability, Reliability and Serviceability

In any SOA based infrastructure Service Response Time (SRT) is one of the key efficiency metrics



SRT Prediction technique



- Represent SRT measurements as time series data
- Apply machine learning technique for prediction in our case ANFIS



Input data structure matters



ANFIS efficiency depends on input data structure

•
$$I'(SRT_{i-2}, SRT_{i-1}, SRT_i)$$

• $I\left(\frac{SRT_i - 2SRT_{i-1} + SRT_{i-2}}{4}, \frac{SRT_i - SRT_{i-1}}{2}, SRT_i\right)$



8 hours forecast for dCap service



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ANFIS vs NARNET, Regression Error Characteristics for MAPE

- AOC 3MF 2.7% Accuracy 97.3%
- AOC 4MF 2.8% Accuracy 97.2%
- AOC 5MF 2.3% Accuracy 97.7%
- AOC NARNET 2.96% Accuracy 97.04%



8 hours forecast for torque service



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ANFIS vs NARNET, Regression Error Characteristics for MAPE

- AOC 3MF 9.6% Accuracy 90.4%
- AOC 4MF 6.4% Accuracy 93.6%
- AOC 5MF 15.4% Accuracy 84.6%
- AOC NARNET 11.8% Accuracy 88.2%



Summary



- ANFIS+SVM allows to predict the service status based on simple monitoring metrics, without digging in the large log files
- Failure prediction is one of the most complex tasks for implementation in Autonomic Computing
- Prediction accuracy with ANFIS up to 8 hours is above 90% and is sufficiently reliable for implementation of automation policies
- SRT prediction with ANFIS is a part of monitoring data analysis process and can be considered as a part of Autonomic Computing
- ANFIS based approach is general and can be applied for other monitoring metrics represented in time series data
- Composite service status identification and prediction with ANFIS based on simple monitoring data is the next step (Composite service multiple services providing a storage or computing functionality)



Thank you for your attention

Backup



BACKUP

Backup2



Data Analysis Method for Failure Prediction is based on Fuzzy Sets Theory – Possibility to map known monitoring terms "OK", "WARNING", "CRITICAL" to monitoring attribute values according to the experts knowledge

Core The system (r ros) for the induct indefinition Core The system experts knowledge Base Detuzzification Interface Fuzzification Interface Fuzzification Interface Fuzzification Interface Fuzzification

Fuzzy Inference System (FIS) for the model identification

For automation of FIS i.e. For adding the "learning" ability - Adaptive Neuro Fuzzy Inference System (ANFIS)

- Automatically identifies the system
- Automatically creates the model for the system
- Based on Feedforward NN with 3 hidden layers learning process uses backpropagation













Load of the System one min. Average and SRT







