Resource Characterization

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VGrADS Vision

Applications and Users





VGrADS Functional Decomposition (so far)



VGrADS Information Services



Resource Characterization

- Abstract description of resources in terms of program accessible attributes
- Quantitative Approach: use <u>automatic</u> statistical methods to capture the dynamics of changing resource characteristics
 - -Monitor data is plentiful and noisy
 - -Must summarize the quantitative behavior of each resource
 - Reduce the complexity associated with using quantitative performance and reliability readings
 - Summaries must be statistically "reliable" to enable effective program-based reasoning and debugging (confidence measures
- Key Questions:
 - Can we make effective quantitative characterizations?
 - Can we deliver the characterizations scalably and fast enough?



Characterization Research

- Leverage previous work:
 - -NWS makes time series predictions for characteristics that are well-modeled as continuously changing levels
 - Network BW and Latency (end-to-end)
 - CPU load
 - Available memory
- New Research: focus on characteristics that do not fit timeseries models well
 - -Resource availability and failure prediction
 - Predicted duration-until-next-failure as a quantitative characterization (Reed+Wolski)
 - Used to schedule checkpoints (Poster by Dan Nurmi)
 - -Batch Queue Wait time prediction
 - New approach to an old problem



Batch Queue Wait Time

- Problem: The <u>vgES</u> and <u>Programming Tools</u> need to know how long individual jobs will wait before they will acquire the resources then need
 - -Perceived execution time is really affected by wait times
 - —Choose the "provisioning" method that best serves the application' needs => VGrADS scheduling
- Goal: Rigorous confidence bounds on the amount of time a specific job will wait in a batch queue before it is scheduled (a cluster or parallel machine.
 - -Statistical nature implies that a quantifiable confidence range is necessary
 - —Need an answer that applies to an individual job Previous work: fi statistical model then predict
 - -Smith, Taylor, Foster (IPDPS, 1999), Downey (IPPS 1997)
 - -Feitelson, http://www.cs.huji.ac.il/~feit/parsched



Modeling and Prediction are Different

- Model fitting
 - -From the distribution, calculating an expectation is possible
 - -Probably not what a user or scheduler needs
 - Mean and variance do not explain what is going to happen to a specific job
- "At most how long will I have to wait before my job runs?"
 The answer is a percentile
- "At most how long will I have to wait before my job runs with 95% confidence?"

- The answer is the 95th percentile

- Goal: estimate the percentiles without explicitly fitting a mod
- <u>Better Goal</u>: estimate percentiles and quantified confidence bounds

- Statistical certainty at specified confidence levels



The Brevik Method

• John Brevik's invention based on Binomial distribution —Probability that exactly j values are below qth quantile is

$$\binom{n}{j} \cdot (1-q)^{n-j} \cdot q^j$$

• Probability that k or fewer values are less than the qth quan

$$\sum_{j=0}^k \binom{n}{j} \cdot (1-q)^{n-j} \cdot q^j$$

- Very robust requiring few sample points (not understood)
- Requires multi-precision arithmetic to calculate because n and can be quite large



How Well Does it Work?

- Examine the batch queue logs that record wait time
- Choose a quantile and a confidence level -0.95 quantile with 95% confidence
- For each job

-Calculate the upper limit on the quantile

-Observe whether job's wait time is less than that limit

• For the entire trace, record the percentage of job wait time that are less than the prediction

-Value should be less than quantile if method is working

- 5 sites and machines (NERSC, LANL, LLNL, SDSC, TACC)
- 9 years (96 through 05)
- 1,200,000+ jobs



Quantifiable Confidence





Capturing Dynamics





Choosing the Best Worst Case





Choosing the Best Number of Processors

Datastar 95% Predictions June 2004, 1-4 and 17-64 Processors





A Batch of Results

- Brevik Method can predict quantiles with specified levels of confidence
 - Must control history adaptively to handle non-stationarity
 - Robust and data frugal enough to work for processor counts too (much harder)
- Combinations of quantiles provide a qualitative way to evalua resources

- If median and 95th percentile are lower, chances are job will start soo

- Quantiles provide a quantitative way to predict possible outcomes
 - 45% chance that a job will start between the median and the 95th percentile
- Possibly New Scheduling Research: Quantitative Contingency Scheduling
 - Build a schedule with contingencies based on quantiles
 - Adjust based on conditional predictions



Delivering the Good News

- Virtualization: construct forecast "snapshot" of resource characteristics for vgES
 - -Use forecasting to cover asynchrony
 - -Use statistical similarity to improve scalability
- Test: <u>NWS Network "Doppler Radar"</u>
- For VGrADS
 - -Replicated name servers (UCSD and Rice)
 - -Replicated data caches (UCSB and Rice)
 - -300 microseconds/forecast over 100mb local ethernet
- VGrADS testbed: <u>The Movie</u>



Conclusions

- New automatic resource characterizations
 - -New approach to batch queue and machine availability
 - -Lead to new scheduling techniques (c.f. Dan Nurmi Poster)
 - -Quantifiable confidence levels
- Result: We provide rigorous bounds on statistical attribute values to <u>vgES</u> and <u>Programming Tools</u>
- New Information System data structures
 - -Scalable and high performance
 - -Provide an instantaneous "picture" of the resources
- Result: Virtualization in the Information System promotes scalability and performance

