

FORMULATING THE GRID RELIABILITY AND PERFORMANCE MODEL

SETTING THE GRID RELIABILITY EQUATIONS AND MATRICIAL MODEL

NYOUG DBA SIG MEETING OCT. 2008 AT ORACLE CORPORATION



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Overview

- **Project Description**
 - **Objective**
 - **Motivation and Expected Results**
- **Conceptual Framework**
 - **Reliability Model**
 - **Performance Model**
 - **Integrating Grid Mathematics**
- **Proposed Model Methodology**
- **Math Applications and Model Analysis**
- **Concluding Remarks**

Speaker Qualifications

- Independent Consultant, ADN
- Speaker at NYOUG meetings
- 24 years of IT experience
- 18 years of Oracle experience, 13 as a DBA
- RMAN experience with Oracle 8i, 9i, 10g, and 11g, since 1999.
- BS Systems Engineering, Universidad del Norte, 1987.
- MS Computer Science, NJIT, 1993
- PhD CIS candidate, NJIT, 1997
- MBA MIS, Montclair State University, 2006
- College Math Professor and former HS Math Teacher Principal.

Objectives

- **Present a mathematical model to customize both grid reliability and performance in terms of:**
 - **Critical Alerts**
 - **Warning Alerts**
 - **Custom Thresholds**
- **Introduce a useful matrix-driven model providing:**
 - **BI multi-dimensional capabilities**
 - **Stochastic Model and random process analysis**
 - **Capabilities to enable time series analysis via ARIMA**

Defining Reliability

- Reliability can be expressed as:

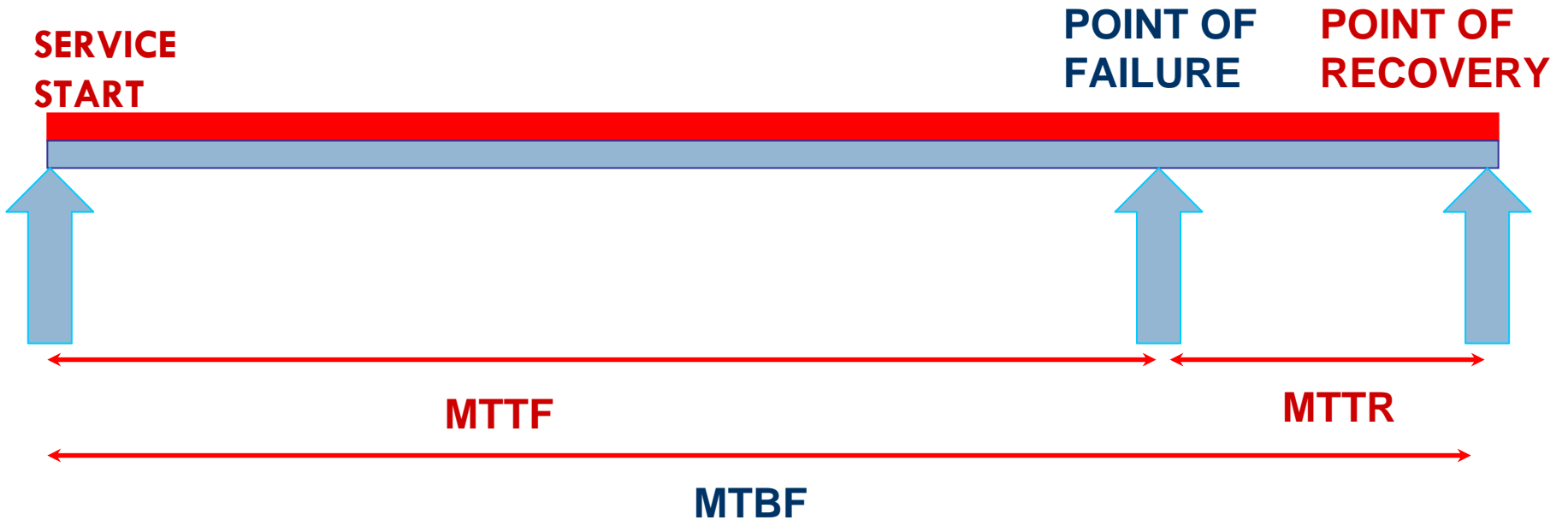
$$R = \frac{MTTF}{MTBF}$$

- And further expanded to:

$$R = \frac{MTTF}{MTTF + MTTR}$$

MTTF is the Mean Time to Failure, and MTBF is the Mean Time Between Failure; and MTBF, and can be decomposed into the Mean Time To Failure plus the Mean Time To Recover (MTTR).

Defining Reliability

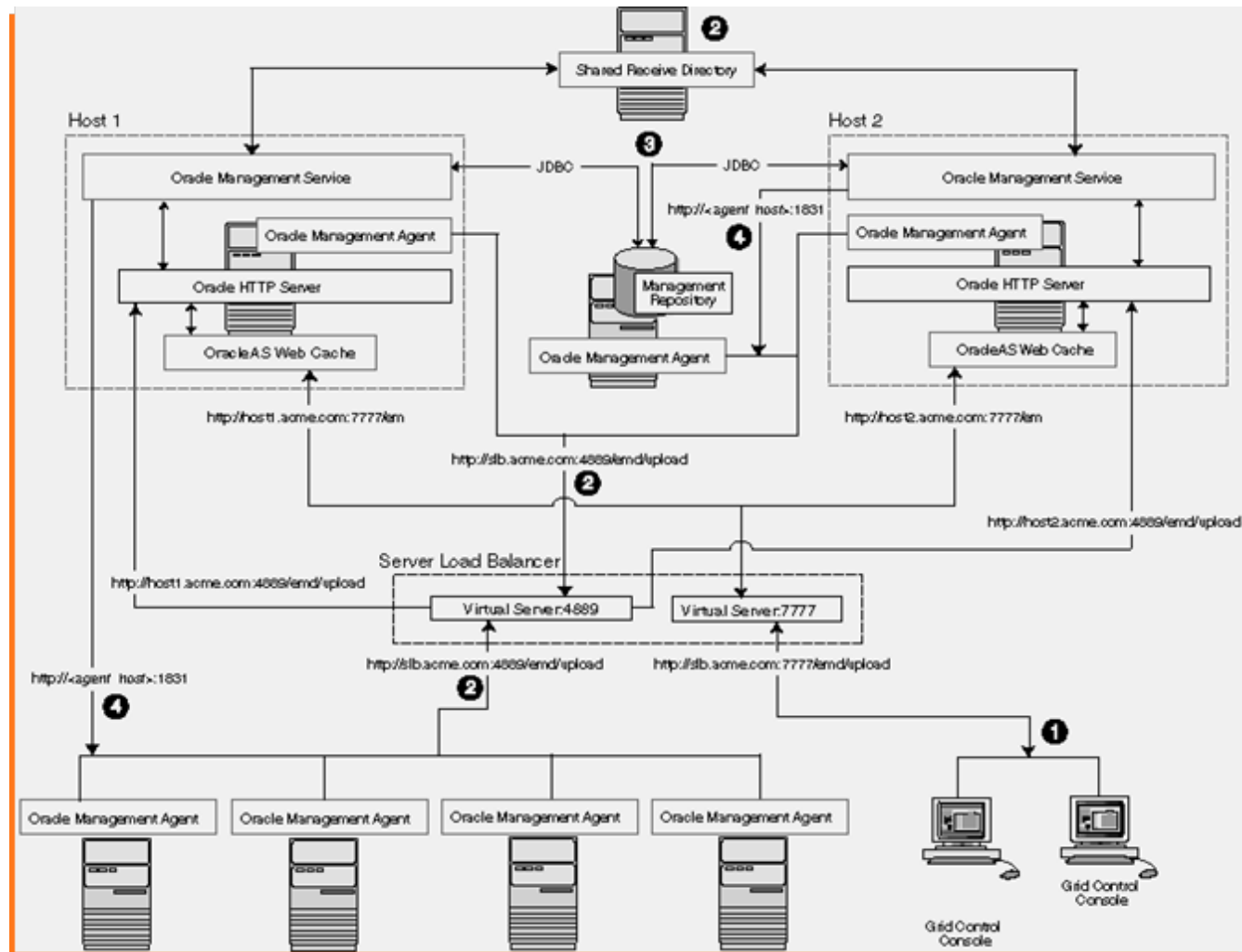


Continuous Reliability Model

$$R = e^{-(\lambda t)} \frac{(\lambda t)^k}{k!}$$

The stochastic model (Poisson process)

Complexity of Grid Infrastructure



General Performance Control Model

- A general performance control model must involve:

- A **Proportionate** approach.
$$Rc = ke^{-(\lambda t)} \frac{(\lambda t)^k}{k!}$$

- An **Integrative** Approach
$$Rc = \int e^{-(\lambda t)} \frac{(\lambda t)^k}{k!} dt$$

- A **Derivative** Approach.
$$Rc = \frac{d}{dt} \left(e^{-(\lambda t)} \frac{(\lambda t)^k}{k!} \right)$$

- So, PID, is the principle to derive performance control based on a continuous reliability model.
- A **Transformative** or **Transformational Approach**, such as Laplace transform, is also suitable for complex systems, where encryption is appropriate.

λ Inter-Arrival Rate Based on Alerts

- **Critical Inter-arrival rate, i.e., based on critical alerts.**

$$\lambda_c = \frac{\# \text{ of Critical Alerts}}{\# \text{ Test Unit}}$$

λ Inter-Arrival Rate Based on Alerts

- **Non-Critical Inter-arrival rate, i.e., based on critical alerts.**

$$\lambda_w = \frac{\# \text{ of Warning Alerts}}{\# \text{ Test Unit}}$$

Weighed Inter-Arrival Rate λ

- **The Model Weighed Inter-arrival Rate Lambda .**

$$\lambda = \frac{n_c \lambda_c + n_w \lambda_w}{n_c + n_w}$$

Adjusting the Inter-arrival rate λ

- **Biased Model Weighed adjusting the Critical Inter-arrival Rate λ (Lambda) .**

$$\lambda = \frac{n_c (k\lambda_c) + n_w \lambda_w}{n_c + n_w}$$

Adjusting the Inter-arrival rate λ

- **Biased Weighed Model** adjusting both the **Critical Inter-arrival Rate λ (Lambda)** and its cardinality .

$$\lambda = \frac{kn_c \lambda_c + n_w \lambda_w}{kn_c + n_w}$$

Weighed Inter-Arrival Rate λ

- **The Model Weighed Inter-arrival Rate Lambda (using a constant for warnings).**

$$\lambda = \frac{n_c \lambda_c + n_w c_w}{n_c + n_w}$$

Adjusting the Inter-arrival rate λ

- **Biased Model Weighed adjusting the Critical Inter-arrival Rate λ (Lambda) and (using a constant for warnings) .**

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Adjusting the Inter-arrival rate λ

- **Biased Weighed Model** adjusting both the **Critical Inter-arrival Rate λ (Lambda)** and its **cardinality** and (using a constant for warnings) .

$$\lambda = \frac{kn_c \lambda_c + n_w c_w}{kn_c + n_w}$$

Aggregate Reliability Model

- **The Aggregate Time-Driven Reliability Continuous Model .**

$$\int_{t_0}^{t_1} \frac{e^{-(\lambda t)} (\lambda t)^k}{k!} dt$$

RAC Reliability Model

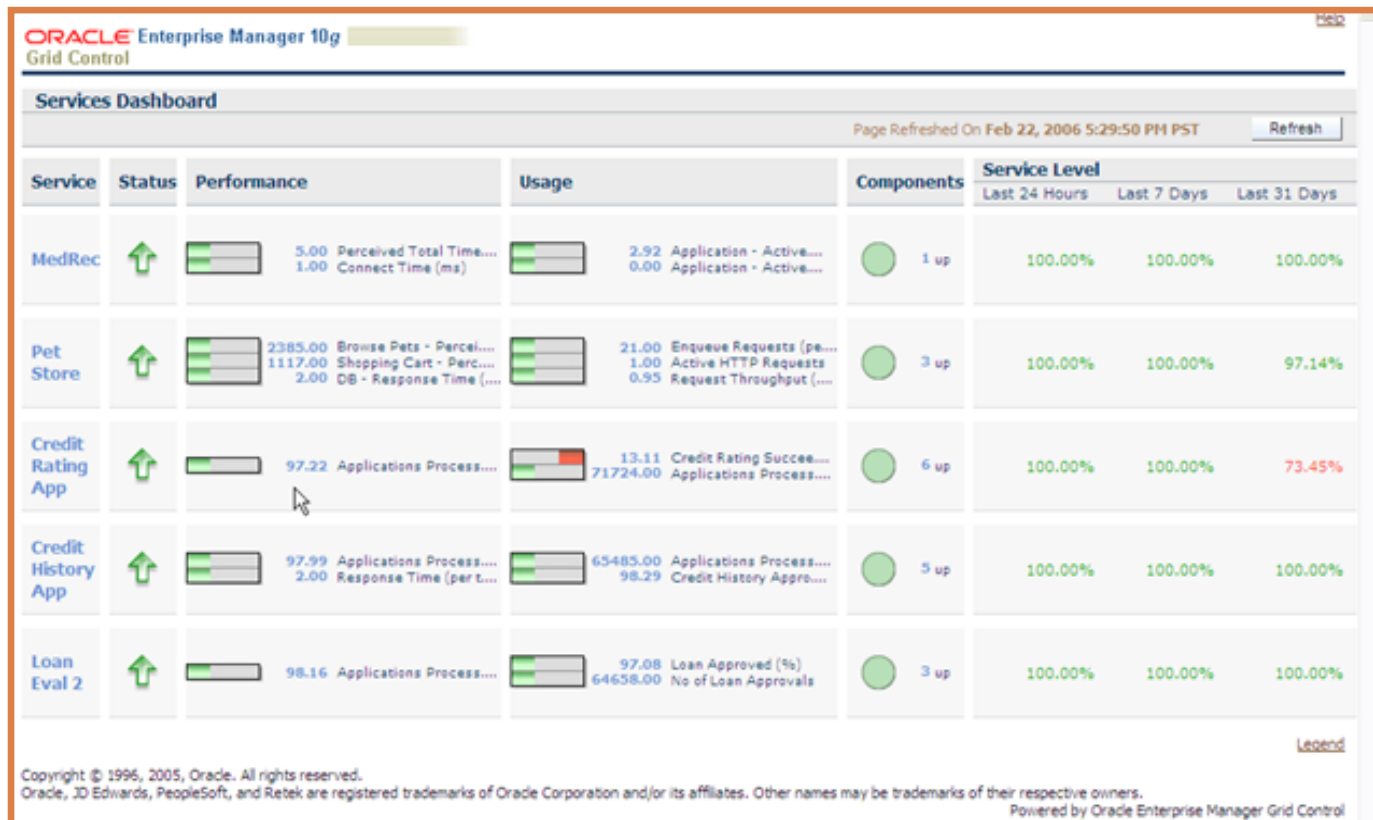
□ RAC Node-based Reliability Expectation.

$$E(\rho_{RAC}) = E\left(\sum_{i=1}^n \frac{MTTF_i}{MTTF_i + MTBF_i}\right)$$

$$E(\rho_{RAC}) = \frac{1}{n} \sum_i \left(\frac{MTTF_i}{MTTF_i + MTBF_i}\right)$$

Grid Reliability Metrics

Actual Grid Reliability Metrics via Service Level Management.



RAC/Grid Pro-Reliability Model

□ Weighed RAC/Grid Inter-Arrival Rate

$$\lambda_{RAC} = \frac{\sum_i^m \sum_j^n n_{c_{ij}} \lambda_{c_{ij}} + \sum_i^m \sum_j^n n_{w_{ij}} \lambda_{w_{ij}}}{\sum_i^m \sum_j^n (n_{c_{ij}} + n_{w_{ij}})}$$

- This can also be presented as the generalized model with cubic features for BI analytic purposes.

RAC/Grid Pro-Reliability Model

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- This can also be presented as the generalized model with cubic features for BI analytic purposes.

General Performance Control Model

- A general performance control model must involve:

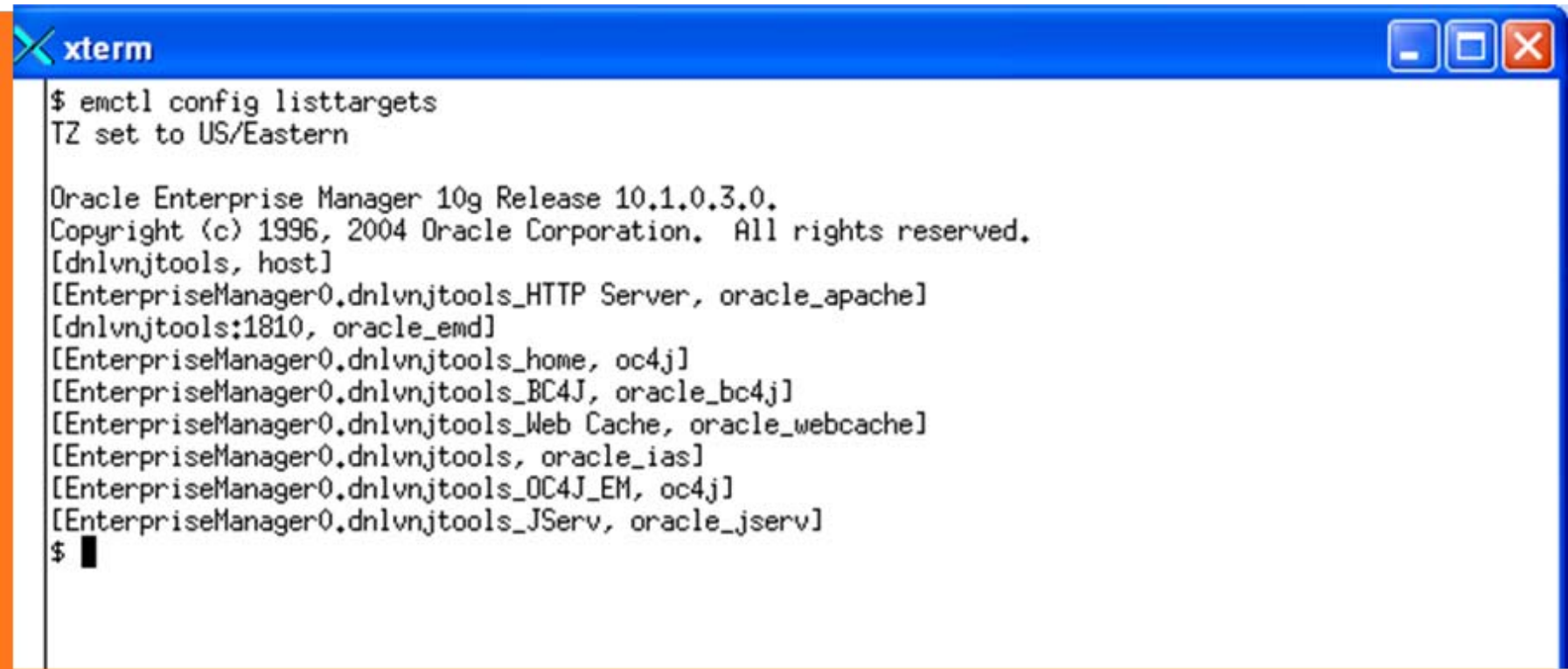
- A **Proportionate** approach.
$$Rc = ke^{-(\lambda t)} \frac{(\lambda t)^k}{k!}$$

- An **Integrative** Approach
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- A **Derivative** Approach.
$$Rc = \frac{d}{dt} \left(e^{-(\lambda t)} \frac{(\lambda t)^k}{k!} \right)$$

- So, PID, is the principle to derive performance control based on a continuous reliability model. As mentioned, a **Transformational** Approach is also possible.

Grid Targets



```
xterm
$ emctl config listtargets
TZ set to US/Eastern

Oracle Enterprise Manager 10g Release 10.1.0.3.0.
Copyright (c) 1996, 2004 Oracle Corporation. All rights reserved.
[dnlvnjtools, host]
[EnterpriseManager0.dnlvnjtools_HTTP Server, oracle_apache]
[dnlvnjtools:1810, oracle_emd]
[EnterpriseManager0.dnlvnjtools_home, oc4j]
[EnterpriseManager0.dnlvnjtools_BC4J, oracle_bc4j]
[EnterpriseManager0.dnlvnjtools_Web Cache, oracle_webcache]
[EnterpriseManager0.dnlvnjtools, oracle_ias]
[EnterpriseManager0.dnlvnjtools_OC4J_EM, oc4j]
[EnterpriseManager0.dnlvnjtools_JServ, oracle_jserv]
$
```


Matricial Models

- Aggregation can be implemented via block matrices.

$$\begin{bmatrix} R & F \\ A & C \end{bmatrix}$$

- The Middleware Block Matrix involving RAC, Application Farms, Applications, and Collaboration Suite Target's Metrics

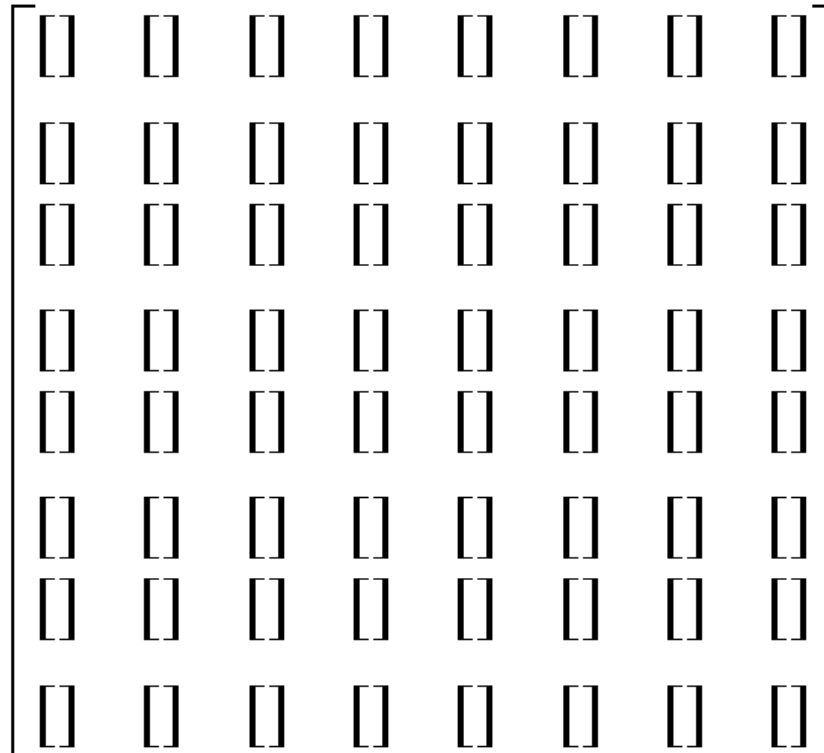
BI Analysis and Forecasting

- **Block Matrix Design, which allows for multi-dimension and Business Intelligence Analysis**

$$\begin{bmatrix} r_{1,1} & r_{1,2} & \dots & r_{1,n/2} & f_{1,1} & f_{1,2} & \dots & f_{1,n/2} \\ r_{2,1} & r_{2,2} & \dots & r_{2,n/2} & f_{2,1} & f_{2,2} & \dots & f_{2,n/2} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ r_{n/2,1} & r_{n/2,2} & \dots & r_{n/2,n/2} & f_{n/2,1} & f_{n/2,2} & \dots & f_{n/2,n/2} \\ a_{1,1} & a_{1,2} & \dots & a_{1,n/2} & c_{1,1} & c_{1,2} & \dots & c_{1,n/2} \\ a_{2,1} & a_{2,2} & \dots & a_{2,n/2} & c_{2,1} & c_{2,2} & \dots & c_{2,n/2} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ a_{n/2,1} & a_{n/2,2} & \dots & a_{n/2,n/2} & c_{n/2,1} & c_{n/2,2} & \dots & c_{n/2,n/2} \end{bmatrix}$$

Block-Based Grid Visualization

- Therefore, one could conceive the grid as a block-based matrix, in general, as displayed



Binomial Model Perspective

Assuming that a grid contains N such block matrices, it is also possible to establish that the probability that a warning or critical alert message reaches one of these blocks over a unit of time is a Bernoulli sequence given by:

$$B(\lambda N, 1/N) = \binom{\lambda N}{k} \left(\frac{1}{N}\right)^k \left(1 - \frac{1}{N}\right)^{\lambda N - k}$$

Where $p = \frac{1}{N}$ is the probability of success, i.e., that an alert reaches a block in the grid over a unit of time.

Binomial Model Perspective

The same Bernoulli trial/sequence perspective can be seen over a period of time as:

$$B((\lambda t)N, 1/N) = \binom{(\lambda t)N}{k} \left(\frac{1}{N}\right)^k \left(1 - \frac{1}{N}\right)^{(\lambda t)N - k}$$

Where $p = \frac{1}{N}$ is the probability of success, i.e., that an alert reaches a block in the grid over a duration of t units of time. The discrete view is consistent with the continuous model previously introduced.

Unified Grid Middleware Model

□ Matrix displaying the Unified Grid Middleware Model

$$\begin{bmatrix} m_{1,1} & m_{1,2} & \dots & m_{1,n/2} & m_{1,(n/2)+1} & m_{1,(n/2)+2} & \dots & m_{1,n} \\ m_{2,1} & m_{2,2} & \dots & m_{2,n/2} & m_{2,(n/2)+1} & m_{2,(n/2)+2} & \dots & m_{2,n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ m_{n/2,1} & m_{n/2,2} & \dots & m_{n/2,n/2} & m_{n/2,(n/2)+1} & m_{n/2,(n/2)+2} & \dots & m_{n/2,n} \\ m_{(n/2)+1,1} & m_{(n/2)+1,2} & \dots & m_{(n/2)+1,n/2} & m_{(n/2)+1,(n/2)+1} & m_{(n/2)+1,(n/2)+2} & \dots & m_{(n/2)+1,n} \\ m_{n/2+2,1} & m_{(n/2)+2,2} & \dots & m_{(n/2)+2,n/2} & m_{(n/2)+2,(n/2)+1} & m_{(n/2)+2,(n/2)+2} & \dots & m_{(n/2)+2,n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ m_{n,1} & m_{n,2} & \dots & m_{n,n/2} & m_{n,(n/2)+1} & m_{n,(n/2)+2} & \dots & m_{n,n} \end{bmatrix}$$

Mathematical Applications

- **Time Series (ARIMA, Smoothing and Moving Averages Methods).**
- **Markov Chains and Bayesian Models**
- **Monte-Carlo Simulation**
- **Structured Matrices, namely, Toeplitz, Hankel, Vandermonde, Cauchy.**

Concluding Remarks

- **Affinity and congruency derived from breaking constraints due to data asymmetry.**
- **Model is statistically and mathematically valid.**
- **Various mathematical applications.**
- **Performance Optimization driven by Proportionate, Integrative, and Derivative PID Reliability Control options.**