VI Workshop PISATEL 2004ERICSSON **S**

A flexible measurement system for supporting Traffic Engineering in MPLS/DiffServ networks

Stefano Giordano

Università di Pisa

Dipartimento di Ingegneria dell'Informazione: Elettronica, Informatica, Telecomunicazioni s.giordano@iet.unipi.it netgroup.iet.unipi.it

A reference scenario

MAID (Multi-Access Inter-Domain) Architecture

Test-bed topology

QoS building blocks

control plane

admission control QoS routing resource reservation

buffer management congestion avoidance queueing and scheduling traffic classification, marking, shaping, policing data plane **metering policy service restoration SLA**

QoS evolution

6**June 1994: rfc1633Integrated Services (IntServ) September 1997: rfc2205 Resource ReSerVation Protocol (RSVP) December 1998: rfc2475Differentiated Services (DiffServ) IntServdifferenziazioneflussiDiffServ Best Effort complessità**

MPLS Support of Differentiated Services **rfc3270, May 2002**

Label Switched Path (LSP)

Universita' di Pisa

```
m20pi-RE0 in
System:
Interface:
            ge-0/2/1
IP:unipi-rtg.pi.garr.net (193.206.136.13)
Max Speed: 200 MBits/s (IP)
BGA:
            50.0 MBits/s (IP)
```
The statistics were last updated Wednesday, 8 December 2004 at 18:13, at which time 'RT-PII-RE0' had been up for 141 days, 4:23:33.

Typical Sampling Interval (and time window)

Max Bits/s: 137.8 Mb/s (275.7%) Average Bits/s: 53.9 Mb/s (107.9%) Current Bits/s: 36.7 Mb/s (73.5%) Max Bits/s: 93.0 Mb/s (186.0%) Average Bits/s: 31.3 Mb/s (62.6%) Current Bits/s: 25.7 Mb/s (51.4%)

'Monthly' Graph (2 Hour Average)

Max Bits/s: 171.3 Mb/s (342.7%) Average Bits/s: 58.5 Mb/s (117.0%) Current Bits/s: 43.6 Mb/s (87.2%) Max Bits/s: 102.5 Mb/s (204.9%) Average Bits/s: 31.4 Mb/s (62.7%) Current Bits/s: 24.9 Mb/s (49.7%)

Control plane functionalities (Multiple Access Signaling – Label Edge Router)

Control plane functionalities (Bandwidth Broker)

LSP management interface

Inter-domain control plane mechanisms via Web I/F

MPLS/DiffServ measurement system

- F. Develop a measurement system to be integrated into a MPLS/DiffServ network for:
	- management purposes (monitoring, statistics, etc…)
	- $\bar{}$ integration into the control plane to automate Resource Allocation, Admission Control (measurement and prediction based) and Traffic **Engineering**
- F. Requirements:
	- **Flexibility**
	- configurability
	- modularity
	- minimize the impact on regular traffic dynamics

Requirements for a client server measurement system

Server-side requirement s:

- **– Me a s u r e me n t s** o f the traffic o ffered to edge routers of a DiffServ/MPLS domain taken over arbitrary and configurable time windows in the past.
- **– Low l aten cy** introduced by the system itself to reduce the unavoidable impact of measurements onto network performance.
- **– Sampling** of net work traffic with accurate timing with low sampling jitter, with no use of busy wait.
- **– Tim e s t a m pin ^g** of traffic samples in order to enable time series processing, including prediction.
- **– Data storage** of per-flow traffic time
series in a proper database.
- **–Remotization**of the measurement system in order to enable remote monitoring and management.
- **– Multiple Client Sup** enable simultaneous operations, such as traffic monitoring, estimation and prediction.
- **– Q uic k** delivery o f information to client/s to prevent that high processing delays impact the relevance of measurements.
- **– I n t e r w o r kin g** with other router subsystems such as routing, MPLS, RSVP-TE daemon, etc.

Client-side requirements:

- **– E s tim a tio ⁿ**. In order to avoid unnecessary CPU load on the border router, the computation of the estimates must be performed in the client side.
- **–Threshold crossing det ection**. When t he traffic load of a flow overcomes apreconfigur e d level t he network control plane must r eact accordingly. The occurrence of a threshold crossing event must be notified as soon as possible to t he con trol plane.

At a glance…

Operational mechanism

Client Interface Manager

$\mathcal{L}_{\mathcal{A}}$ Client Interface Manager (CIM)

 application that manages the Client Interface connected to the associated domain Edge Routers (coordinates modules, mechanisms and events in the client side)

Application: monitoring/management

Application: monitoring/management

\mathbb{R}^2 **RRDTool** - produced graphs

Application: monitoring/management

- \mathcal{L} Tool developed b y the Networks Group of the University of Pisa for a MPLS/DiffServ network traffic monitoring
	- $\mathcal{L}_{\mathcal{A}}$ **Graphs for each LSP/PHB pair are available (measured and predicted** values)

Lorenz Equations

Lorentz attractor

Attracting sets

 $\bar{}$ For every $\underline{x}(t_0) {=} \underline{x}_0$ it is possible to find the solution $\phi_\mathrm{t}(\underline{x}_0)$

called **flow** of the system

Strange Attractors

Chaotic predictor

 $\mathcal{L}_{\mathcal{A}}$ The basic idea in adopting a chaotic predictor is that a windowed portion of the samples of a stochastic process changes due to an inner choatic behavior represented by the nonlinear map:

Steps to construct the prediction alg orithm

 \mathbb{R}^n Consider the time series as obtained by sampling a continuous time function as

 $\overline{}$ The state vector at time n is then:

$$
x_n = x(n\tau)
$$

nen:
$$
\underline{x}_n = \begin{pmatrix} x(n\tau) \\ x((n-1)\tau) \\ \vdots \\ x((n-m+1)\tau) \end{pmatrix}
$$
 Dimension m

- $\overline{}$ \blacksquare The parameter τ is also called **delay time**
- \mathcal{C} **Select a proper interpolation function** f_{γ} **such that** $\underline{x}_{n+1} = f_N(\underline{x}_n)$

 \mathcal{C} \blacksquare Notice that, with this choice of state space, the future state \underline{X}_{n+1} has only **sthe first component unknown!**

RBFP - Radial Basis Functions Predictor (2)

 $\overline{}$ The form of the radial basic functions φ:R+*→*R is the following: $\phi(r) = (r^2 + c^2)^{-\beta}, \quad \beta > -1 \quad e \quad \beta \neq -1$

 c and β are costants

 $\mathcal{L}_{\mathcal{A}}$ **Coefficients** λ_i are determined through the knowledge of the past state evolution by imposing:

$$
\hat{x}_{n_i+1} = \sum_{j=1}^{k} \lambda_j \phi \left(\left\| \underline{x}_{n_i} - \underline{x}_{n_j} \right\| \right) + \mu \qquad \text{and} \qquad \sum_{j=1}^{k} \lambda_j = 0
$$

which leads to the linear system of $(k+1)$ equations (to be solved by numerical techniques):

$$
\begin{pmatrix}\n\phi_{11} & \cdots & \phi_{1k} & 1 \\
\vdots & \ddots & \vdots & \vdots \\
\phi_{k1} & \cdots & \phi_{kk} & 1 \\
1 & \cdots & 1 & 0\n\end{pmatrix}\n\begin{pmatrix}\n\lambda_1 \\
\vdots \\
\lambda_k \\
\mu\n\end{pmatrix} =\n\begin{pmatrix}\nx_{n_1+1} \\
\vdots \\
x_{n_k+1} \\
0\n\end{pmatrix}
$$

Application to Network Traffic

m. **Videoconference traffic (prototypal codec)**

m. **Time series: number of ATM cells transmitted during a frame period (1/25 sec.)**

- ٠ **Average bit rate 1.668 M bps**
- ٠ **Peak-to-mean ratio: 4.8**
- ٠ **Frame per second: 25**
- ٠ **Numb er of frames: 48496**

Automated mechanism of RBFP produces:

- ٠ **Embedding dimension m*=3**
- $\mathcal{L}_{\mathcal{A}}$ **Delay-time = 40 msec**
- × **Number of neighbors** $k = m^*+1 = 4$
- **Value of c=0.28**
- **Number of samples N=3000**

RBFP Predictor performance: Video Sequence

RBFP Predictor performance: Video Sequence

A simple case of linear prediction

Application: Traffic Engineering

Application: Traffic Engineering

Application: Traffic Engineering

a $\left(\text{primary and eventually backup path} \right)$

From ITU-D SG2 Question 16/2 "Teletraffic Engineering Handbook" – Chap.1

Teletraffic theory is an inductive discipline. From the observation of real systemswe establish theoretical models fromwhich we derive parameters which can be compared with the corresponding observation of real systems If there is agreement he model has been validated. If not, then we have to elaborate the model further.

From ITU-D SG2 Question 16/2 "Teletraffic Engineering Handbook" – Chap.1

From ITU-D SG2 Question 16/2 "Teletraffic Engineering Handbook" – Chap.1

• **Mathematical Model**

(inexpensive; no time consuming, general)

• **Simulation Model**

(expensive; simulation model is not general: every individual case must be simulated; confidence intervals; etc)

Theory of discrete state Stochastic Processes

Discrete-event simulation: Trace Driven(inputs from collected data) orArtificial Inputs from statistical distributions

Prototypes: field-trials, real and artificial generators, analyzers, etc

42

(even more expensive, specific, time and resource consuming)

