

A Signal Analysis of Network Traffic Anomalies

Paul Barford

with Jeffery Kline, David Plonka, Amos Ron

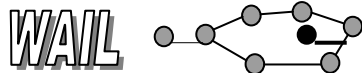
University of Wisconsin – Madison

Fall, 2002



Overview

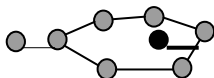
- **Motivation:** Anomaly detection remains difficult
- **Objective:** Improve understanding of traffic anomalies
- **Approach:** Multiresolution analysis of data set that includes IP flow, SNMP and an anomaly catalog
- **Method:** Integrated Measurement Analysis Platform for Internet Traffic (IMAPIT)
- **Results:** Identify anomaly characteristics using wavelets and develop new method for exposing short-lived events



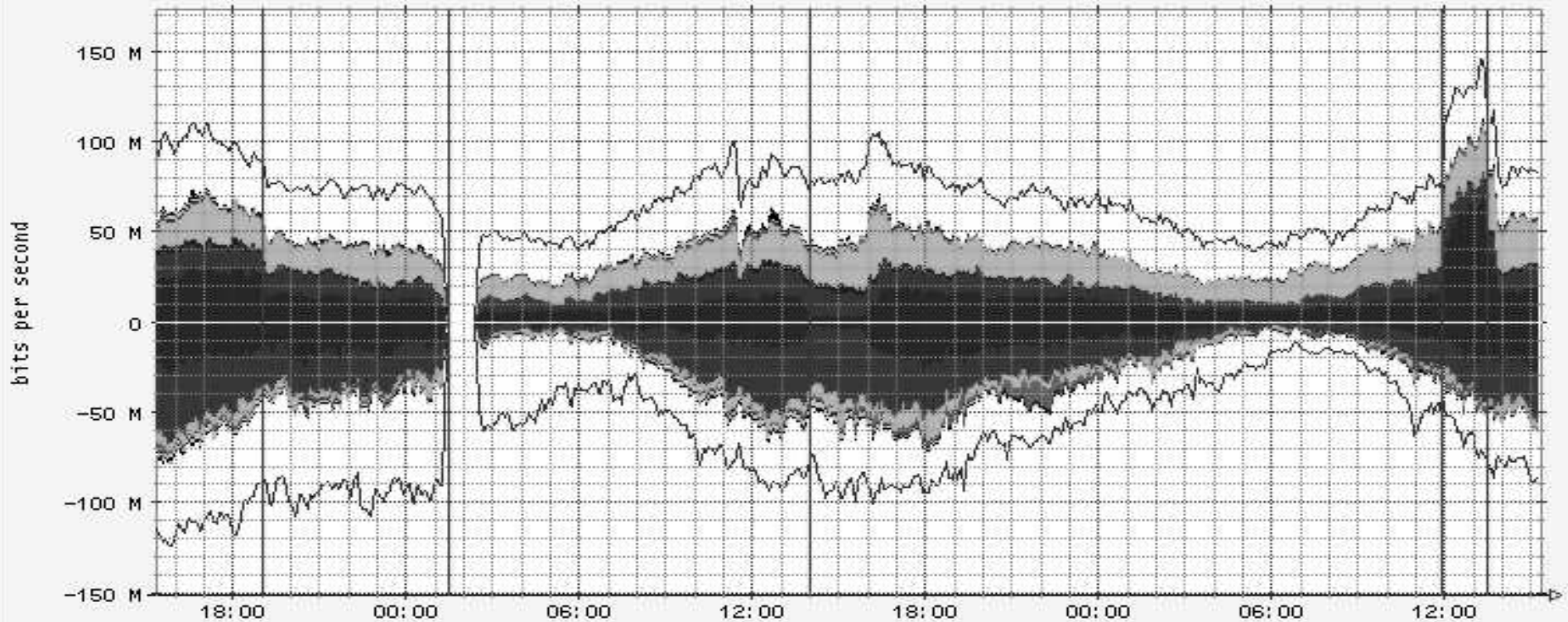
Our Data Sets

- Consider anomalies in IP flow and SNMP data
 - Collected at UW border router (Juniper M10)
 - Archive of ~6 months worth of data (packets, bytes, flows)
 - Includes catalog of anomalies (after-the-fact analysis)
- Group observed anomalies into four categories
 - **Network anomalies (41)**
 - Steep drop offs in service followed by quick return to normal behavior
 - **Flash crowd anomalies (4)**
 - Steep increase in service followed by slow return to normal behavior
 - **Attack anomalies (46)**
 - Steep increase in flows in one direction followed by quick return to normal behavior
 - **Measurement anomalies (18)**
 - Short-lived anomalies which are not network anomalies or attacks

WAIL

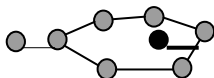


UW-Madison Well Known Services, +out/-in, 8-FEB-2001 -> 10-FEB-2001



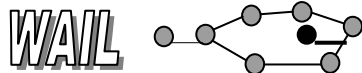
■ Napster*	19.4% Out	15.7% In		
■ HTTP src	+	■ HTTP dst	15.1% Out	25.3% In
■ FTP DATA src	+	■ FTP DATA dst	24.8% Out	9.9% In
■ MCAST	0.1% Out	0.0% In		
■ NNTP src	+	■ NNTP dst	0.8% Out	1.0% In
■ Realserver	0.8% Out	1.4% In		
■ SMTP src	+	■ SMTP dst	0.7% Out	1.0% In
■ ICMP	0.1% Out	0.1% In		
■ Scour exchange	0.0% Out	0.0% In		
Other	38.3% Out	45.5% In		

- TOTAL
- 2001/02/08 1902 applied 33.6Kb/s limit on ResNet-to-world Napster data flows (other Resnet-to-world remains 100Kb/s)
- 2001/02/09 0130 platform Catalyst/ATM problem caused measurement outage (50 mins)
- 2001/02/09 1400 Napster.com outage/problems? (this was independently observed by other FlowScan sites)
- 2001/02/10 1155 removed ResNet rate-limits
- 2001/02/10 1326 routed ResNet through RiverStone router (reactivating rate-limits)



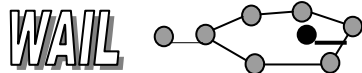
Multiresolution Analysis

- Wavelets provide a means for describing time series data that considers both *frequency* and *time*
 - Powerful means for characterizing data with sharp spikes and discontinuities
 - Using wavelets can be quite tricky
- We use tools developed at UW which together make up IMAFIT
 - FlowScan software
 - The IDR Framenet software



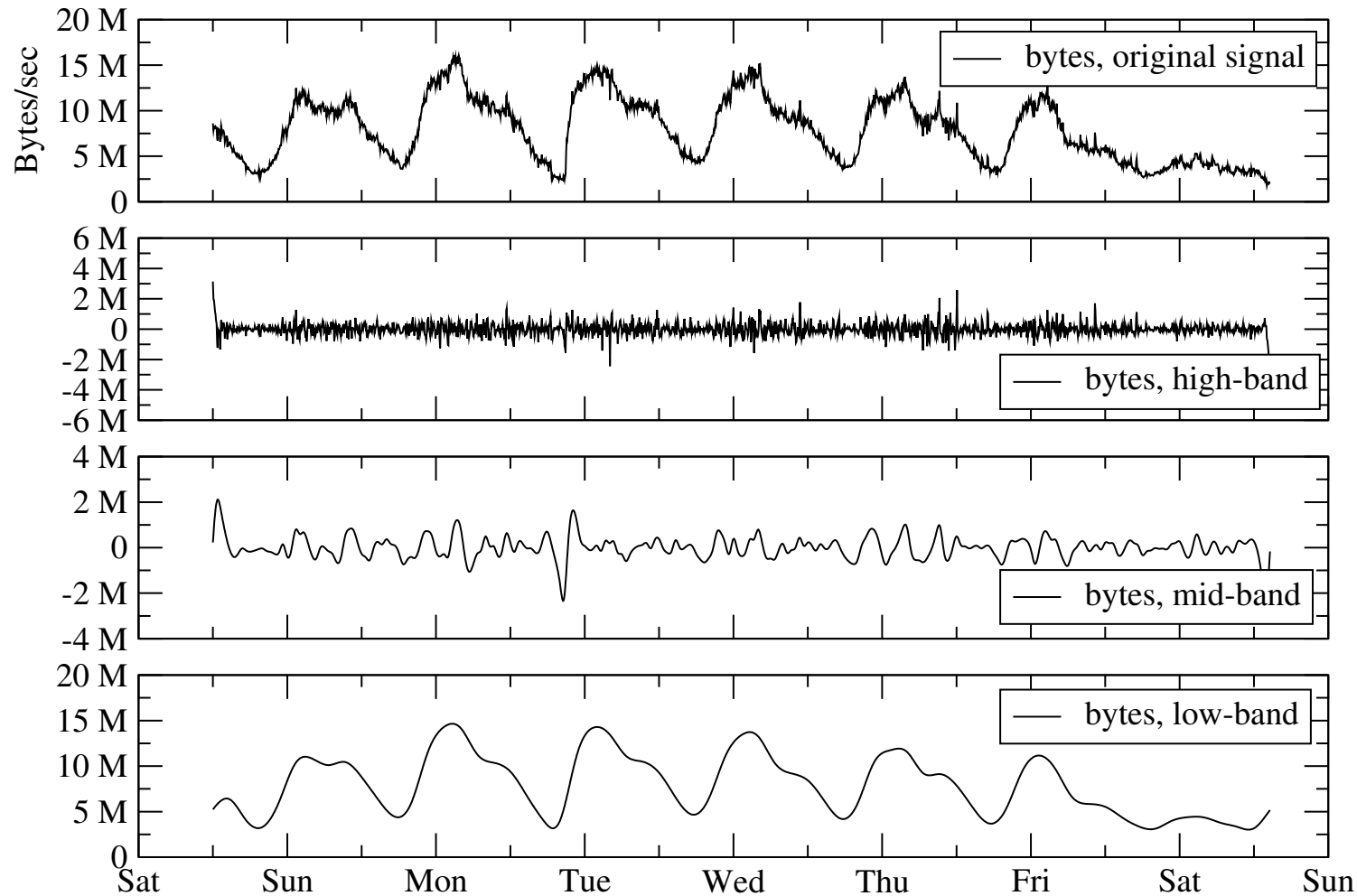
Our Wavelet System

- After evaluating different candidates we selected a wavelet system called Pseudo Splines(4,1) Type 2.
 - A *framelet* system developed by Daubechies *et al.* '00
 - Very good frequency localization properties
- Three output signals are extracted
 - **Low Frequency (L)**: synthesis of all wavelet coefficients from level 9 and up
 - **Mid Frequency (M)**: synthesis of wavelet coefficients 6, 7, 8
 - **High Frequency (H)**: synthesis of wavelet coefficients 1 to 5



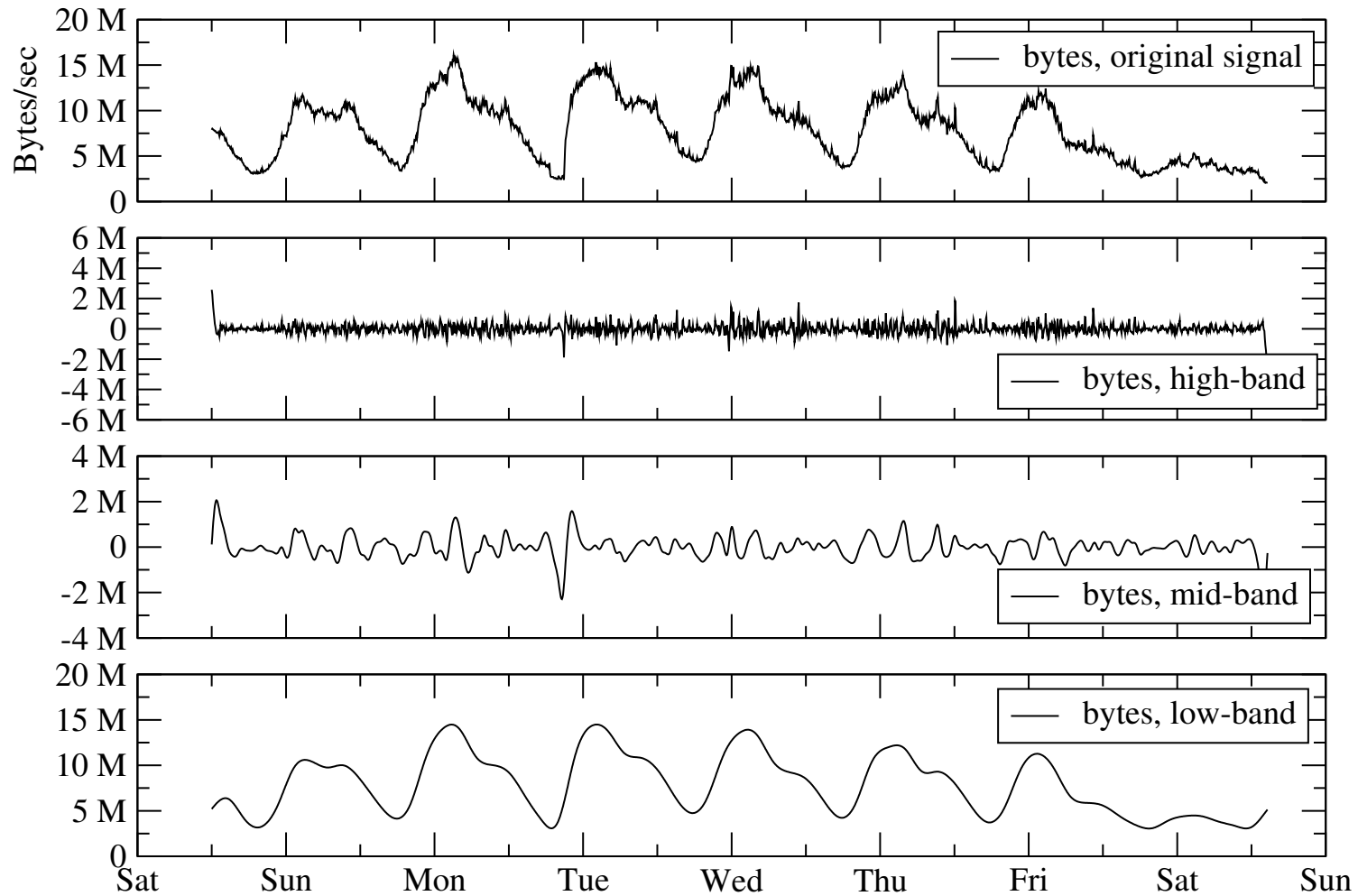
Ambient IP Flow Traffic

One Autonomous System to Campus, Inbound, 2001-DEC-16 through 2001-DEC-23



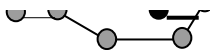
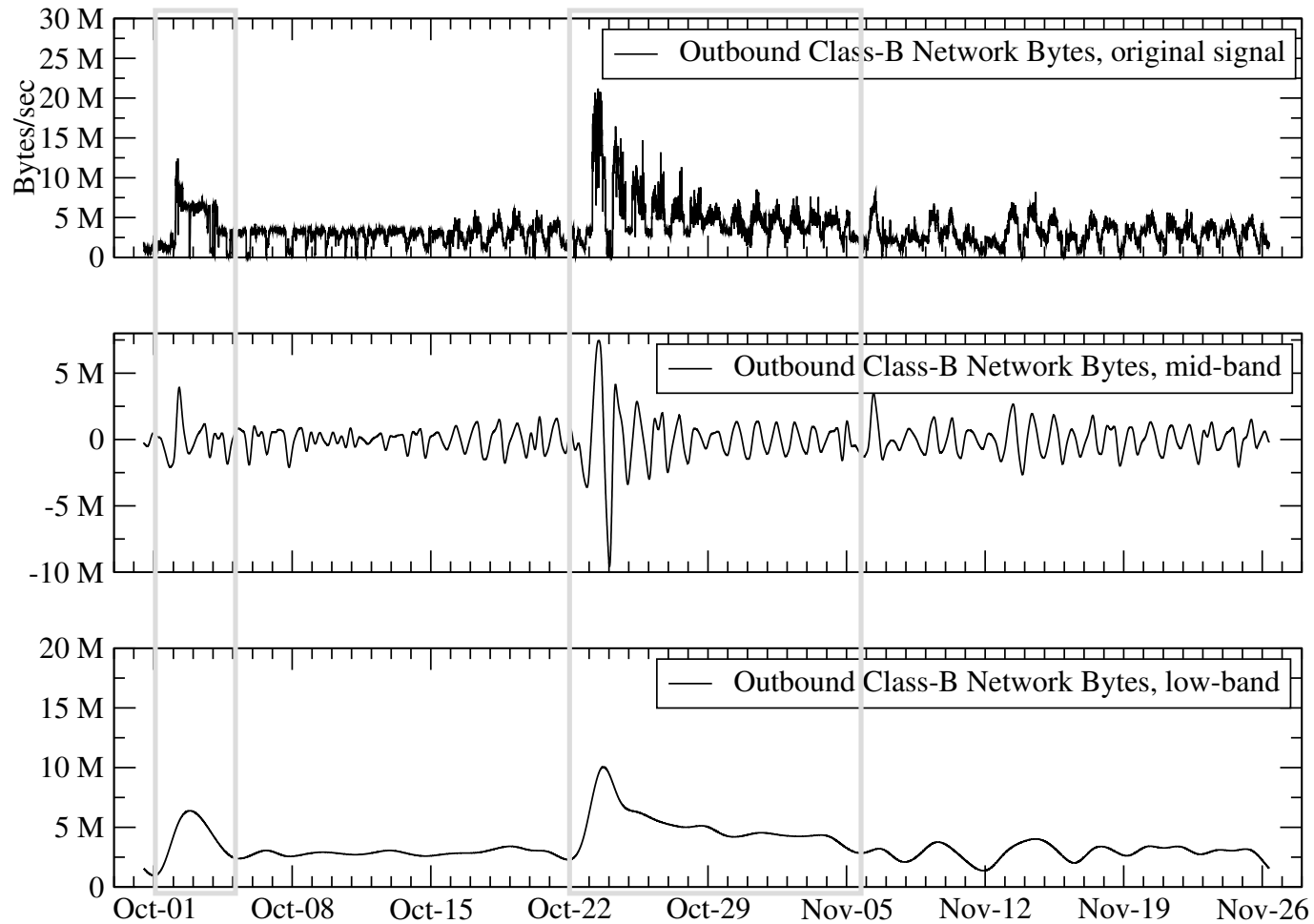
Ambient SNMP Traffic

One Interface to Campus, Inbound, 2001-DEC-16 through 2001-DEC-23



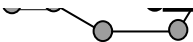
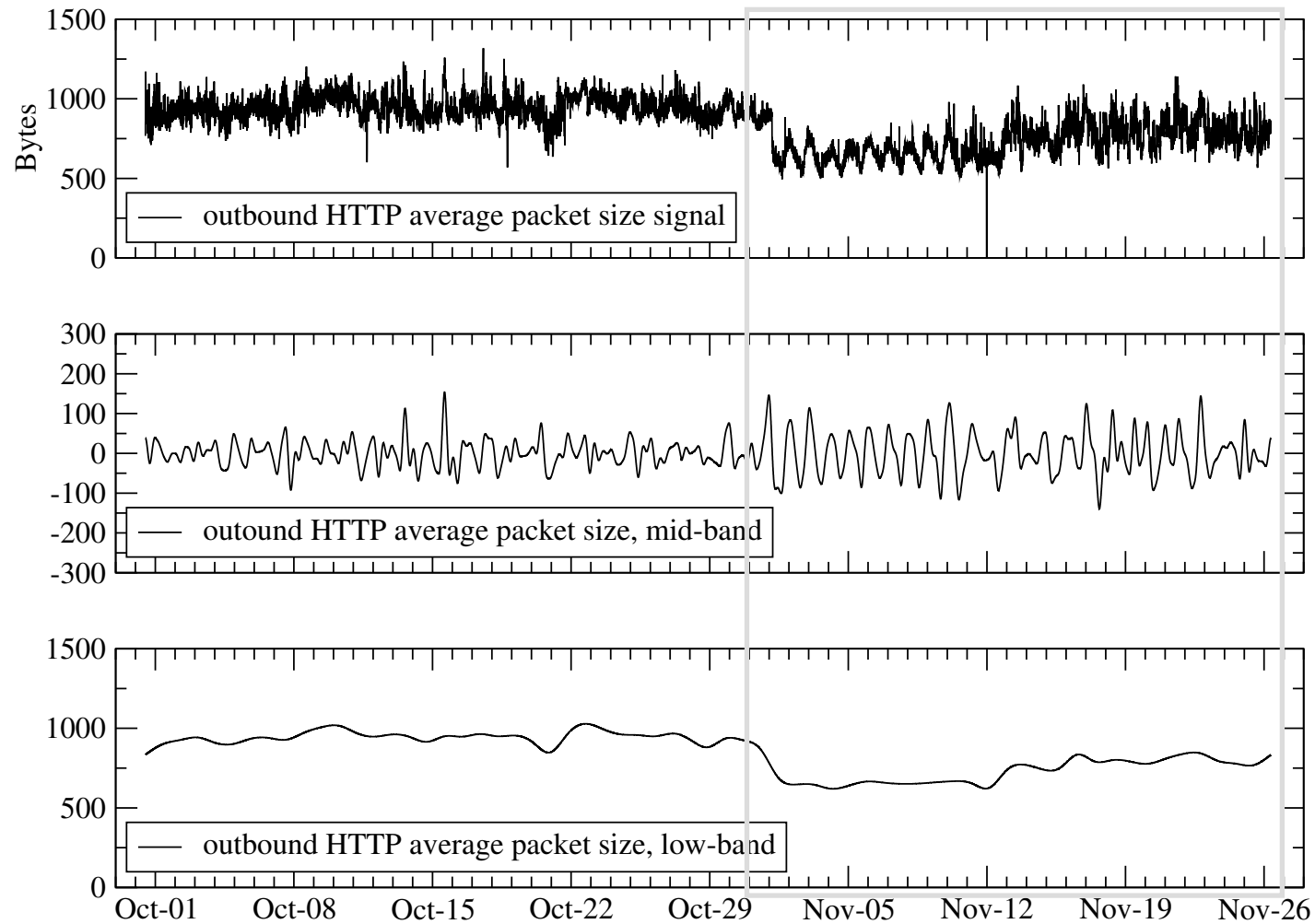
Byte Traffic for Flash Crowd

Class-B Network, Outbound, 2001-SEP-30 through 2001-NOV-25



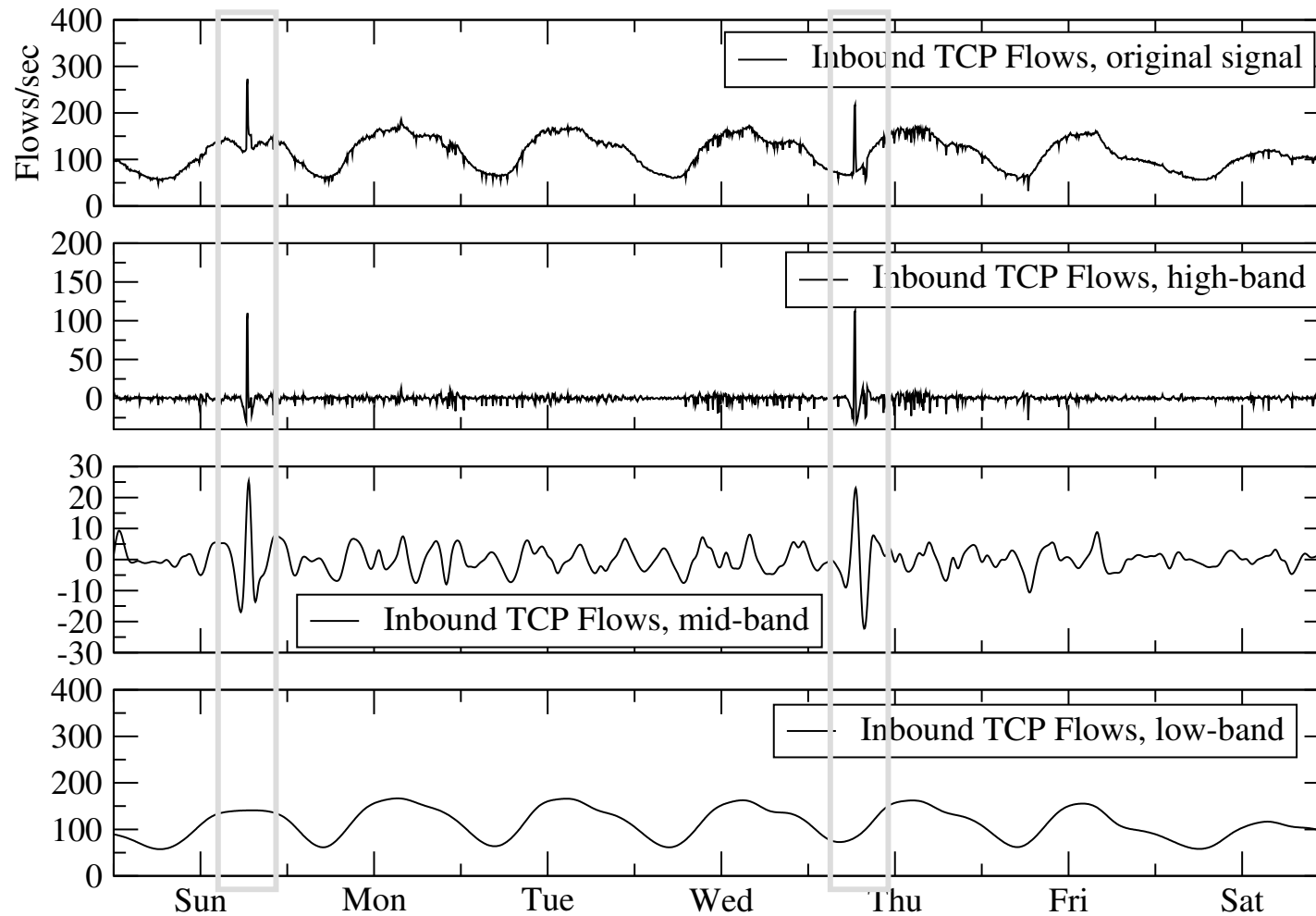
Average Packet Size for Flash Crowd

Campus HTTP, Outbound, 2001-SEP-30 through 2001-NOV-25



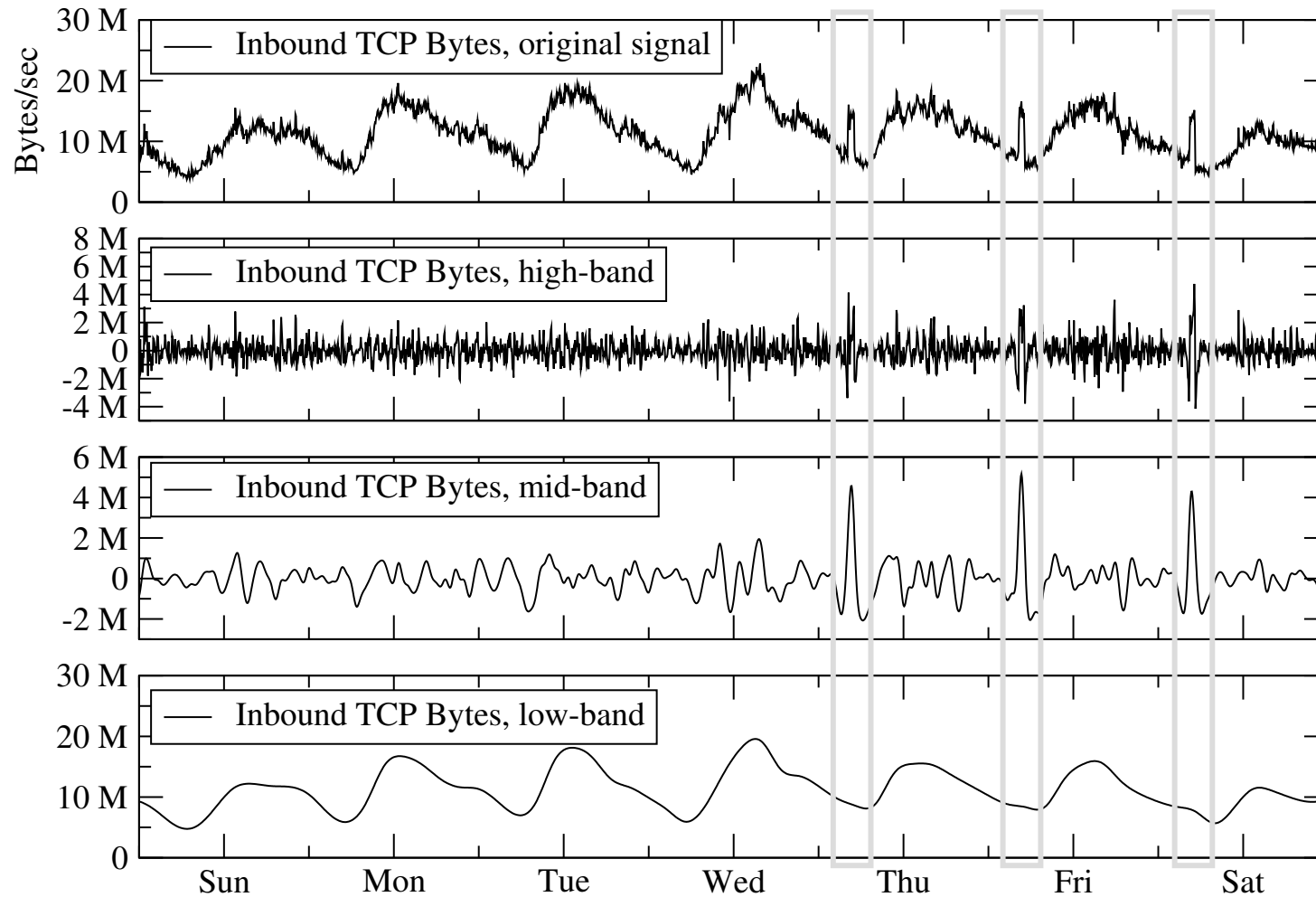
Flow Traffic During DoS Attacks

Campus TCP, Inbound, 2002-FEB-03 through 2002-FEB-10



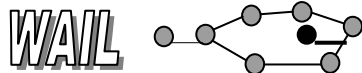
Byte Traffic During Measurement Anomalies

Campus TCP, Inbound, 2002-FEB-10 through 2002-FEB-17



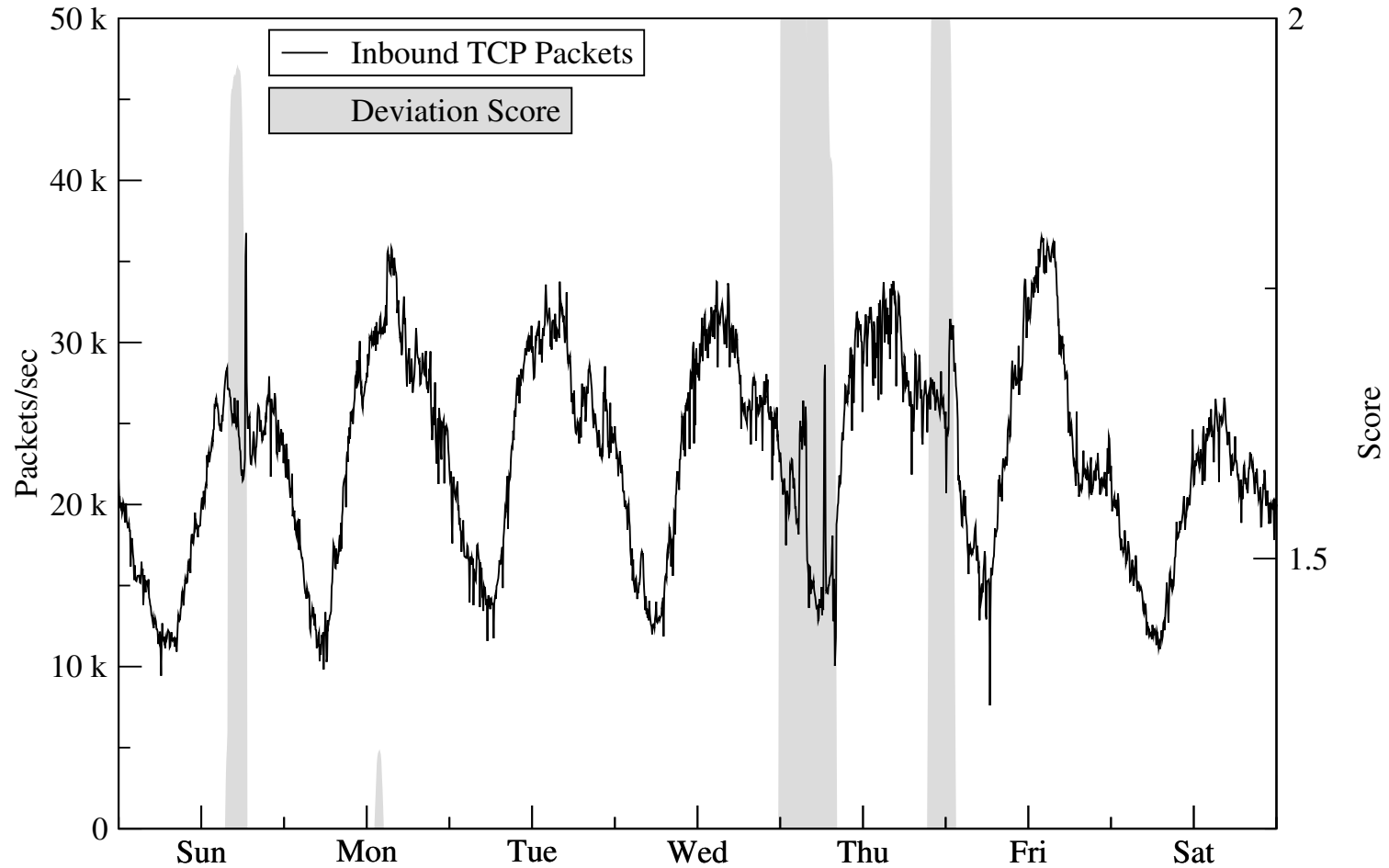
Anomaly Detection via Deviation Score

- Short-lived anomalies can be identified automatically based on variability in H and M signals
 1. Compute local variability (using specified window) of H and M parts of signal
 2. Combine local variability of H and M signals (using a weighted sum) and normalize by total variability to get deviation score V
 3. Apply threshold to V then measure peaks
- Analysis shows that V peaks over 2.0 indicate short-lived anomalies with high confidence
 - We threshold at $V = 1.25$ and set window size to 3 hours

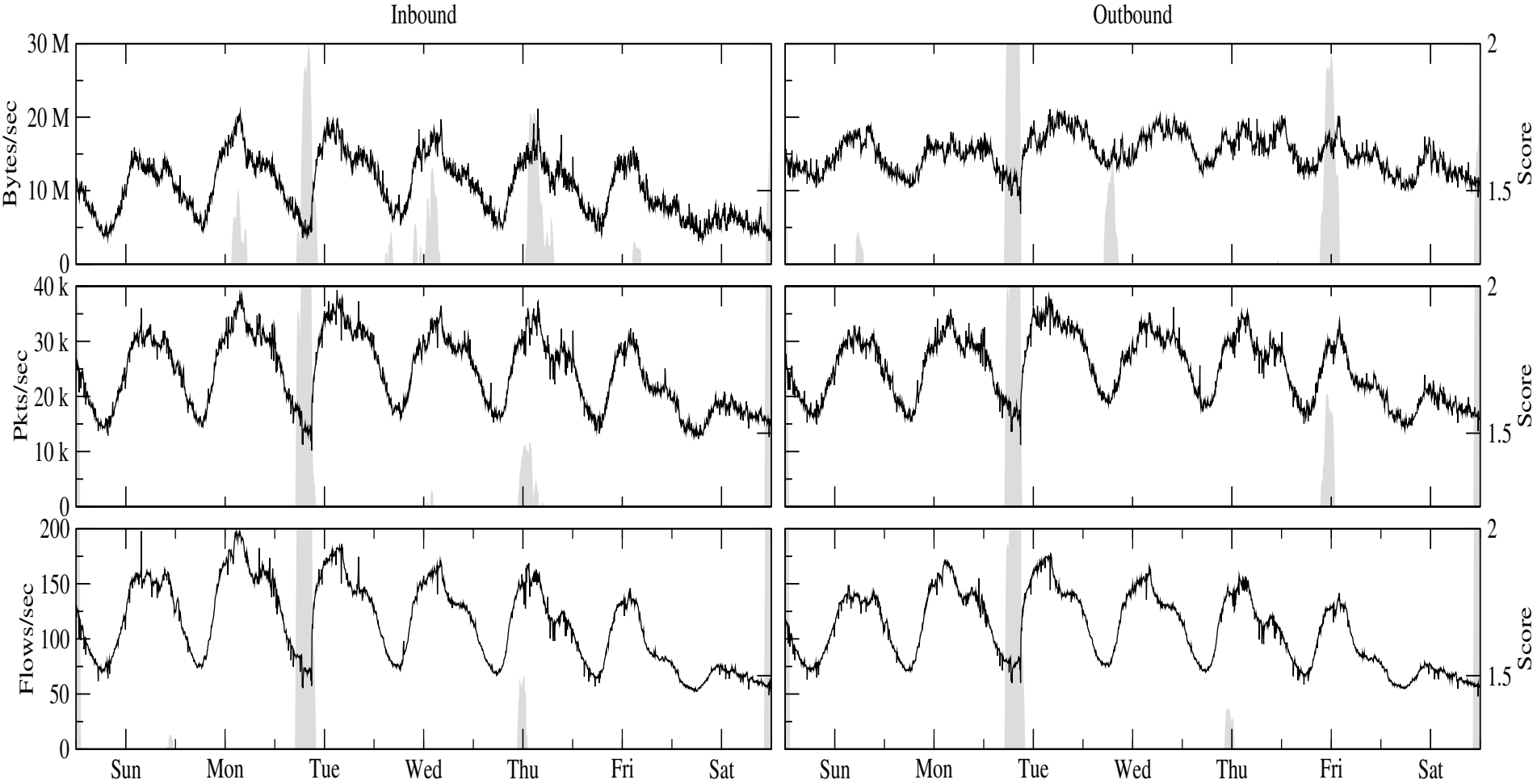


Deviation Score for Three Anomalies

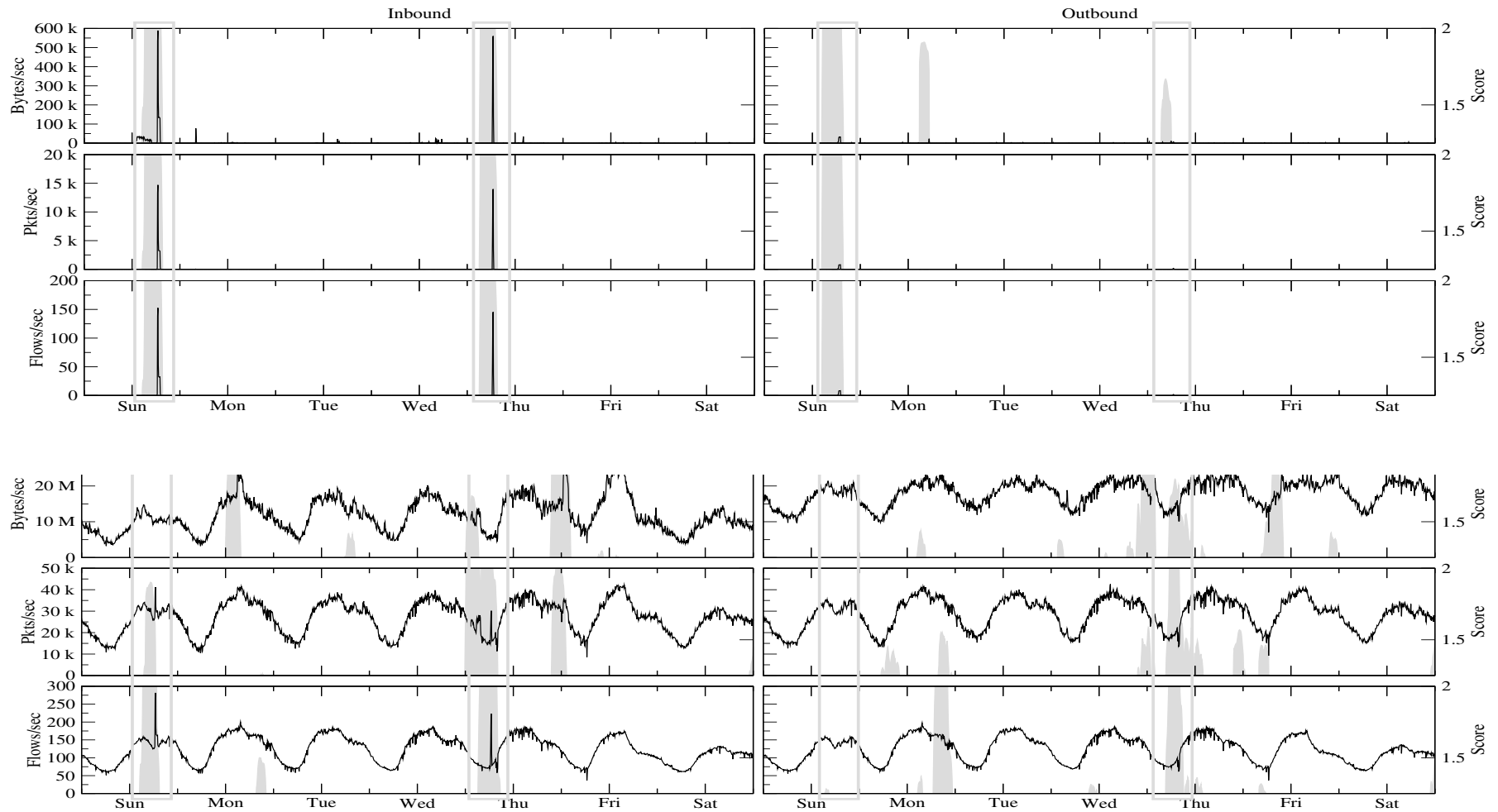
Campus TCP, Inbound, 2002-FEB-03 through 2002-FEB-10



Deviation Score for Network Outage

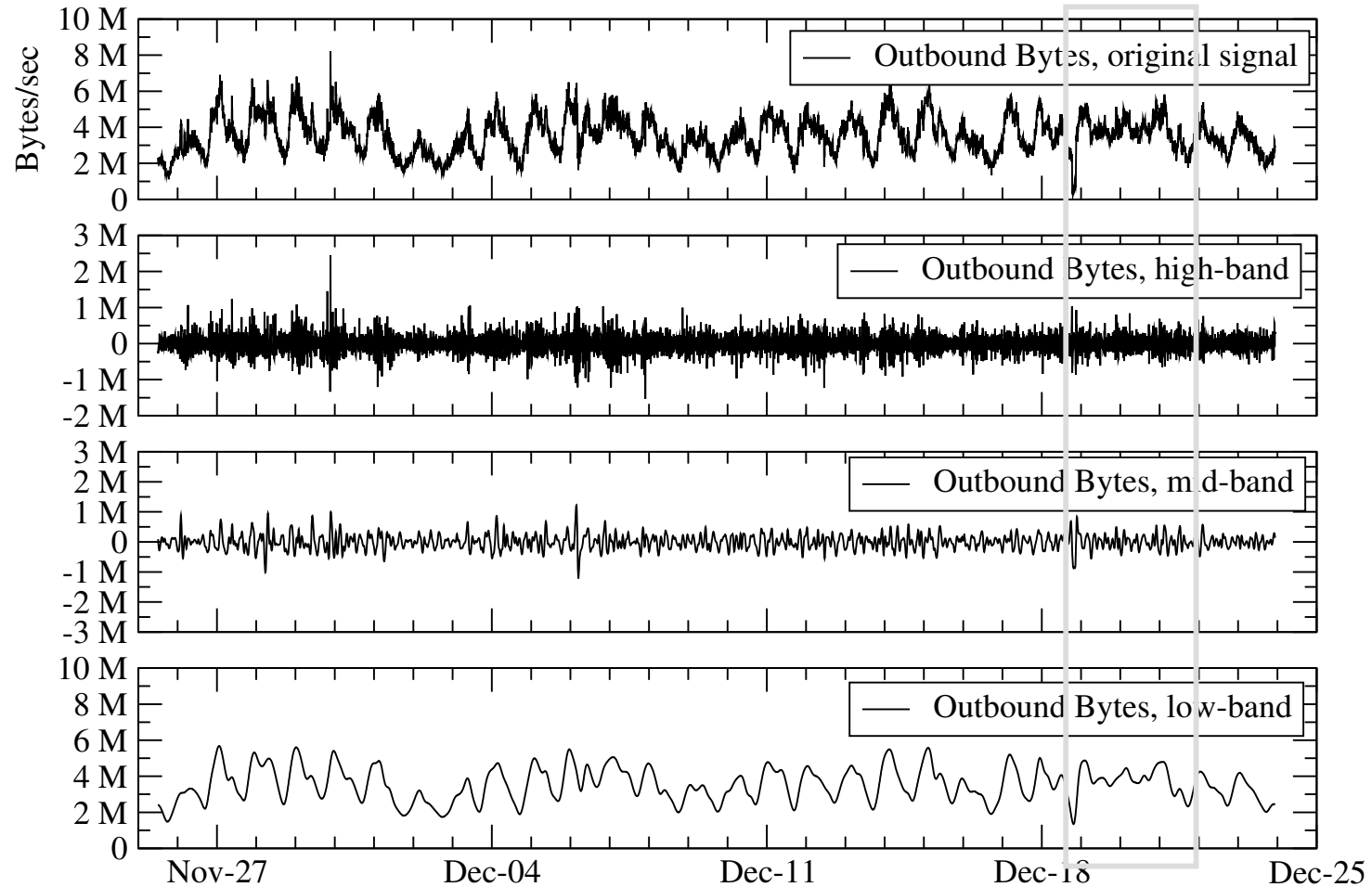


Anomalies in Aggregate Signals



Hidden Anomalies in Low Frequency

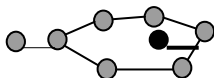
Class-B Network, Outbound, 2001-NOV-25 through 2001-DEC-23



Deviation Score Evaluation

- How effective is deviation score at detecting anomalies?
 - Compare versus set of 39 anomalies
 - Set is unlikely to be complete so we don't treat false-positives
 - Compare versus *Holt-Winters Forecasting*
 - Time series technique
 - Requires some configuration
- Holt-Winters reported many more positives and sometimes oscillated between values

Total Candidate Anomalies	Candidates detected by Deviation Score	Candidates detected by Holt-Winters
39	38	37



Conclusion and Next Steps

- We present an evaluation of signal characteristics of network traffic anomalies
 - Using IP flow and SNMP data collected at UW border router
 - IMAPIT developed to apply wavelet analysis to data
 - Deviation score developed to automate anomaly detection
- Results
 - Characteristics of anomalies exposed using different filters and data
 - Deviation score appears promising as a detection method
- Future
 - Development of anomaly classification methods
 - Application of results in (distributed) detection systems

