

# A Signal Analysis of Network Traffic Anomalies

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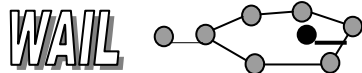
University of Wisconsin – Madison

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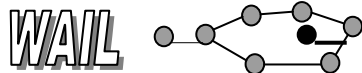
# Overview

- **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

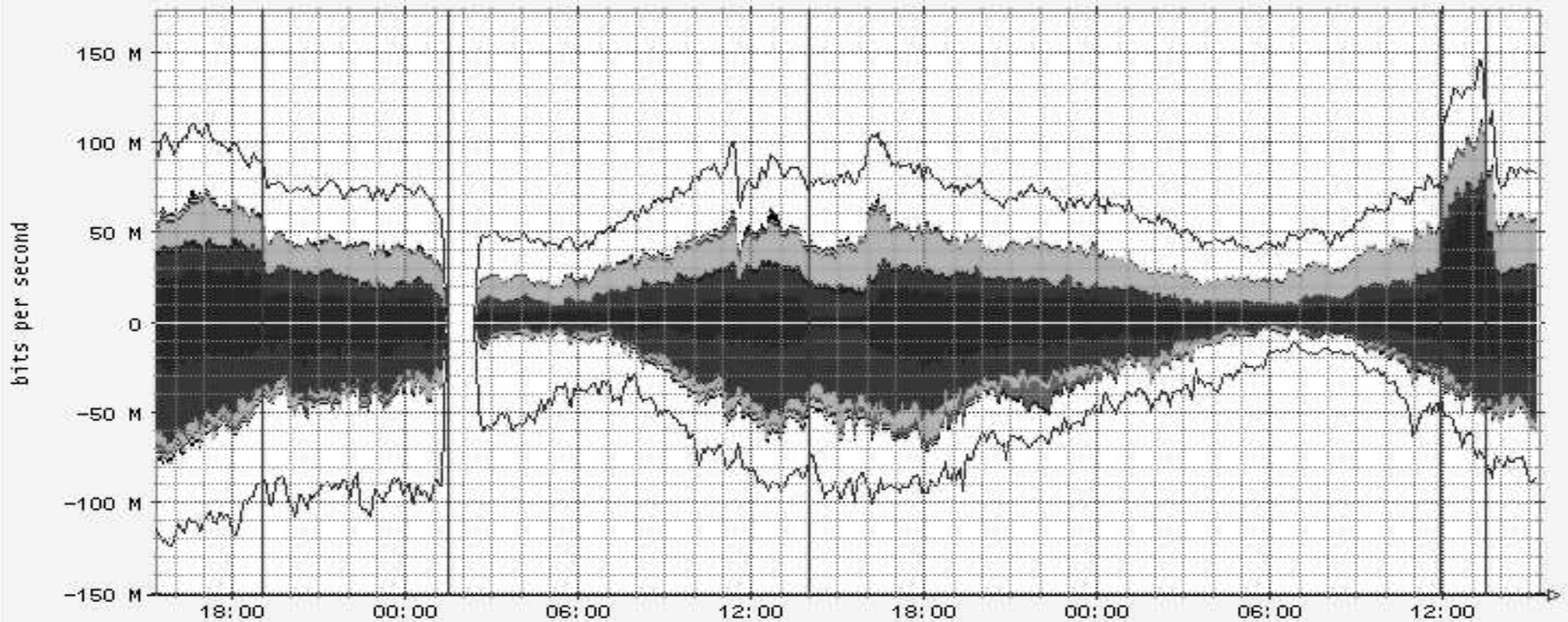


# Our Data Sets

- 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

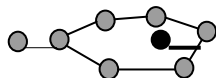


UW-Madison Well Known Services, +out/-in, 8-FEB-2001 -> 10-FEB-2001



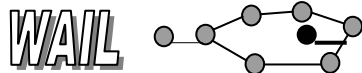
■ Napster*	19.4% Out	15.7% In		
■ HTTP src + ■ HTTP dst	15.1% Out	25.3% In		
■ FTP DATA src + ■ FTP DATA dst	24.8% Out	9.9% In		
■ MCAST	0.1% Out	0.0% In		
■ NNTP src + ■ NNTP dst	0.8% Out	1.0% In		
■ Realserver	0.8% Out	1.4% In		
■ SMTP src + ■ SMTP dst	0.7% Out	1.0% In		
■ ICMP	0.1% Out	0.1% In		
■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



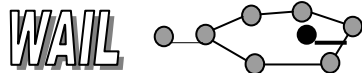
# Multiresolution Analysis

- 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



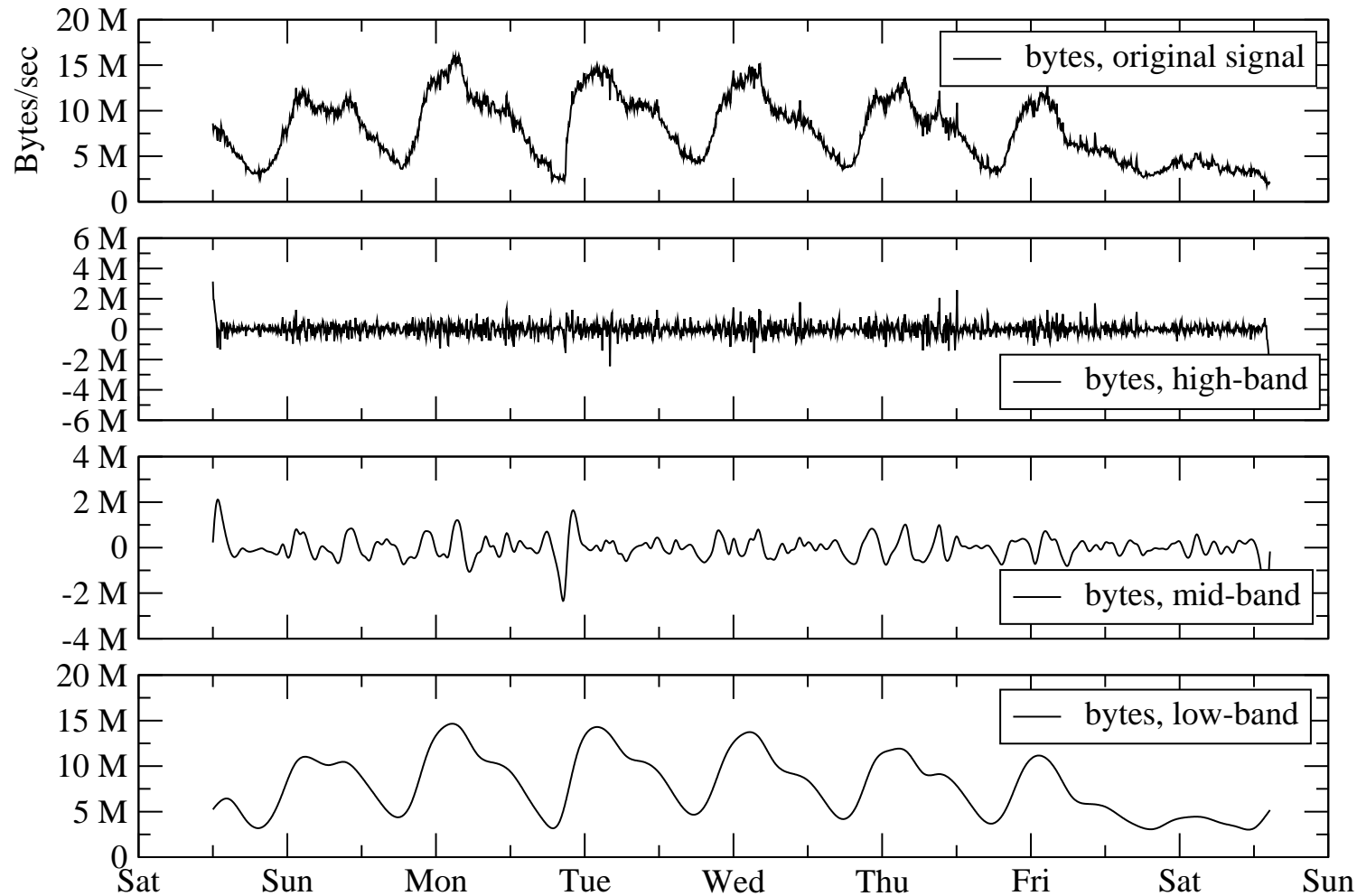
# Our Wavelet System

- 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



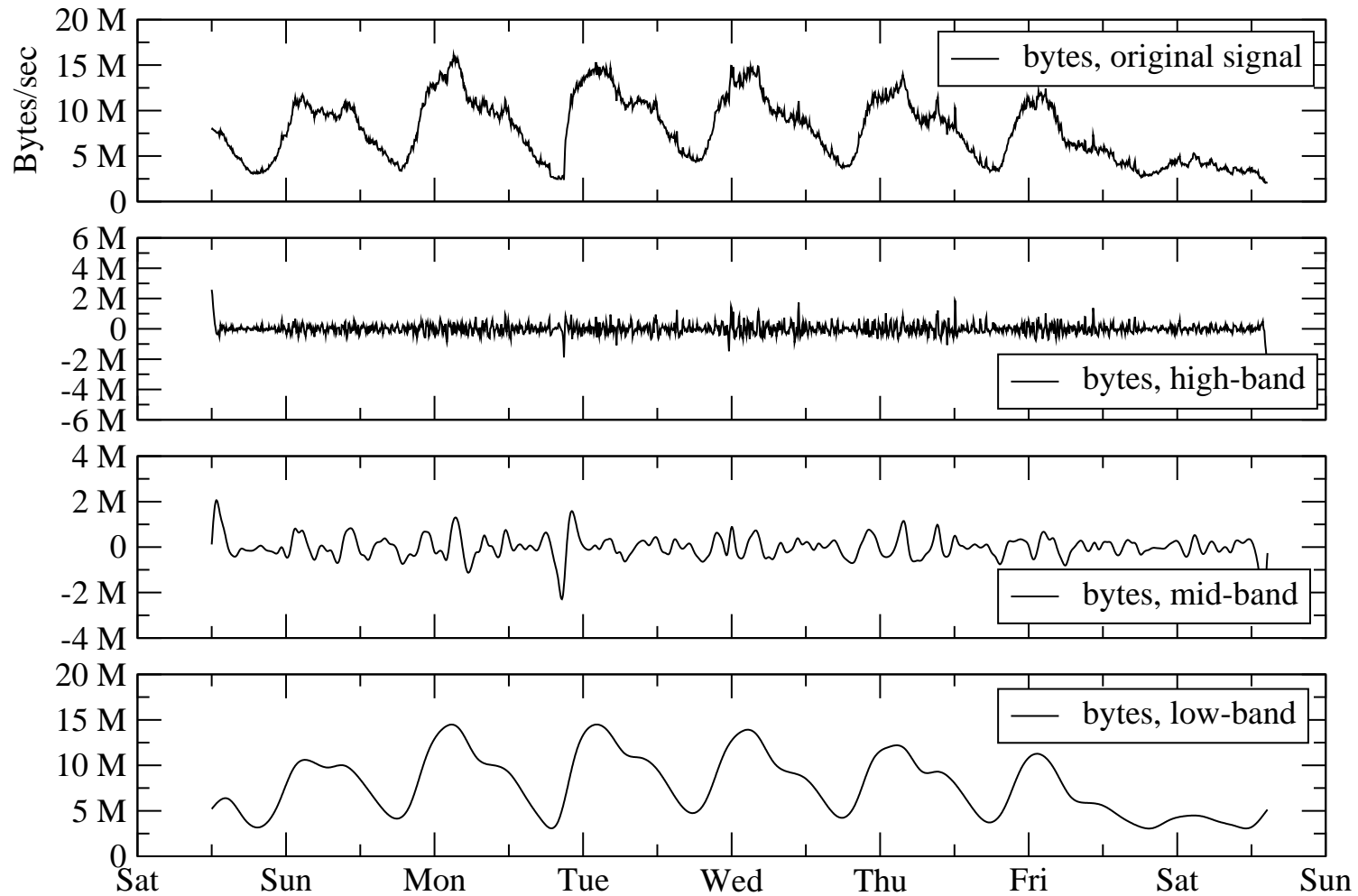
# Ambient IP Flow Traffic

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



# Ambient SNMP Traffic

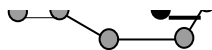
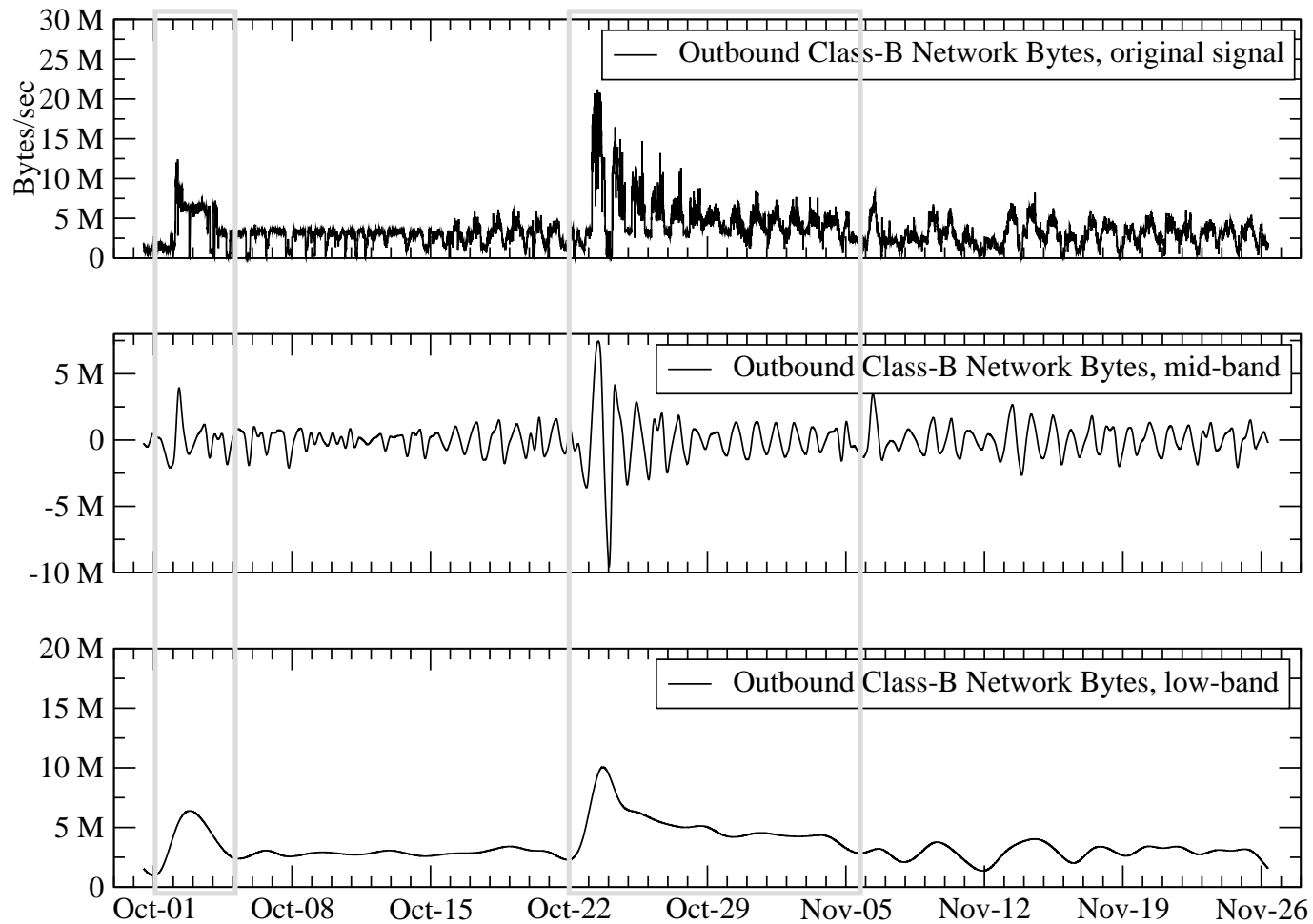
One Interface to Campus, Inbound, 2001-DEC-16 through 2001-DEC-23





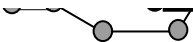
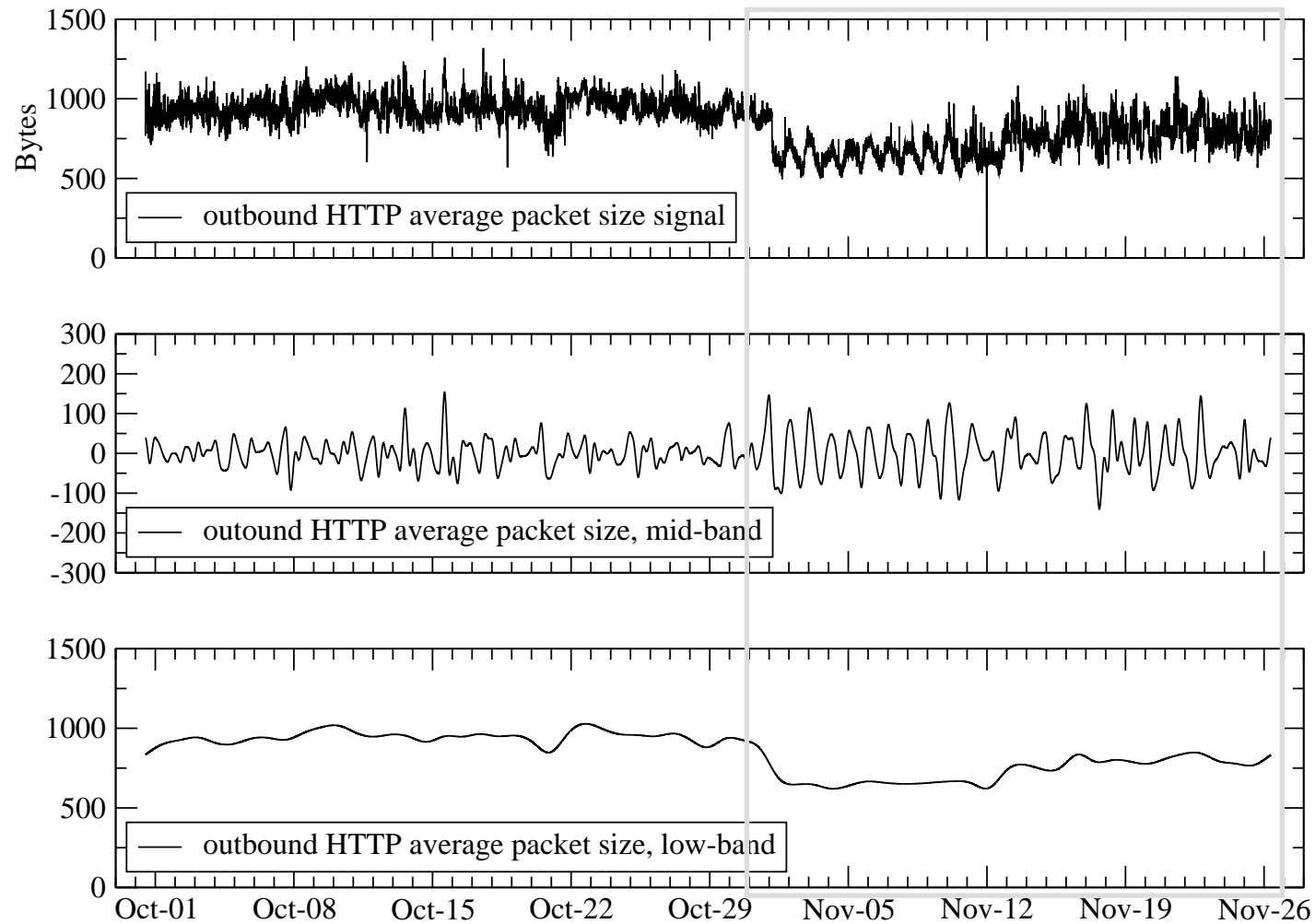
# Byte Traffic for Flash Crowd

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



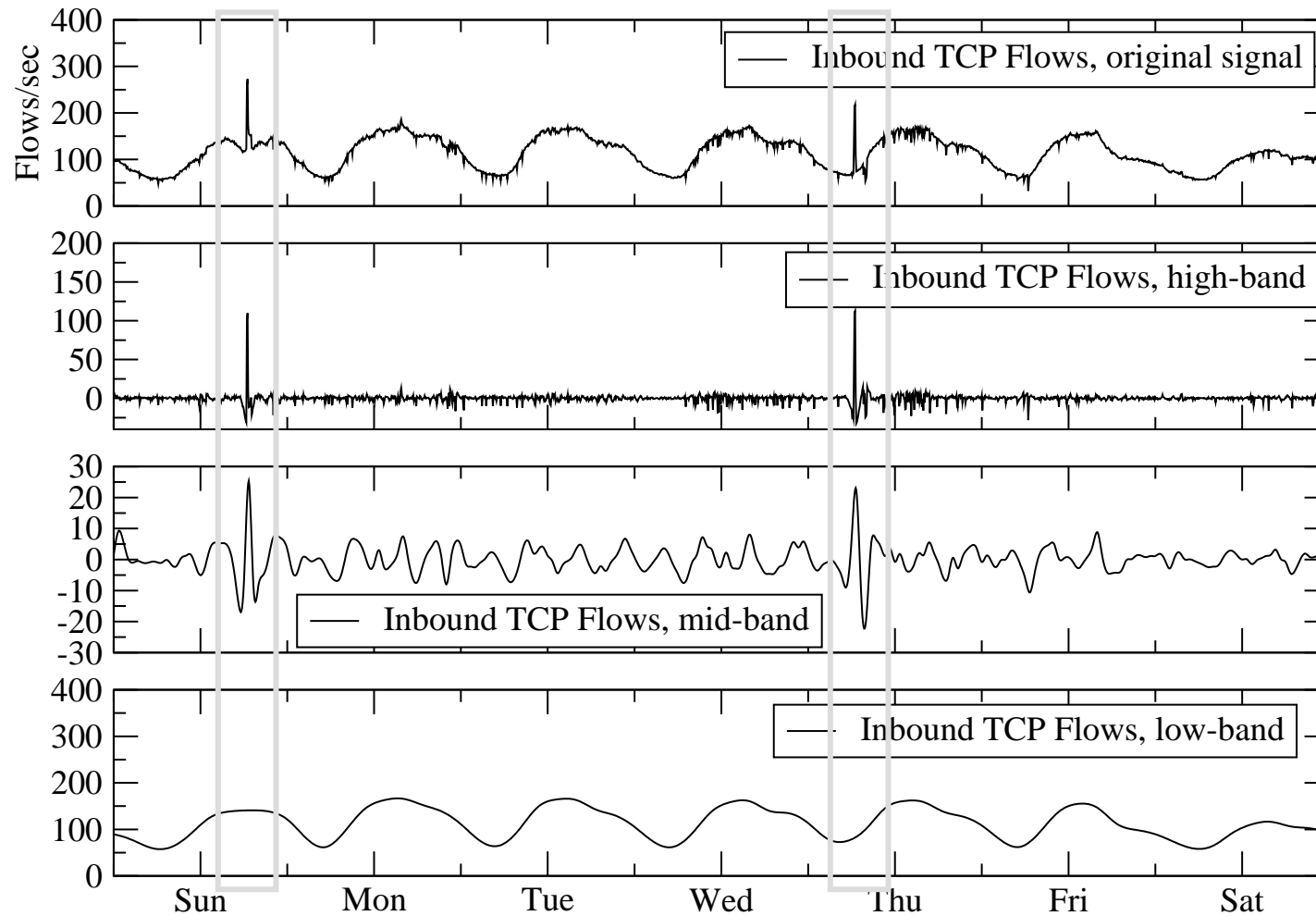
# Average Packet Size for Flash Crowd

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



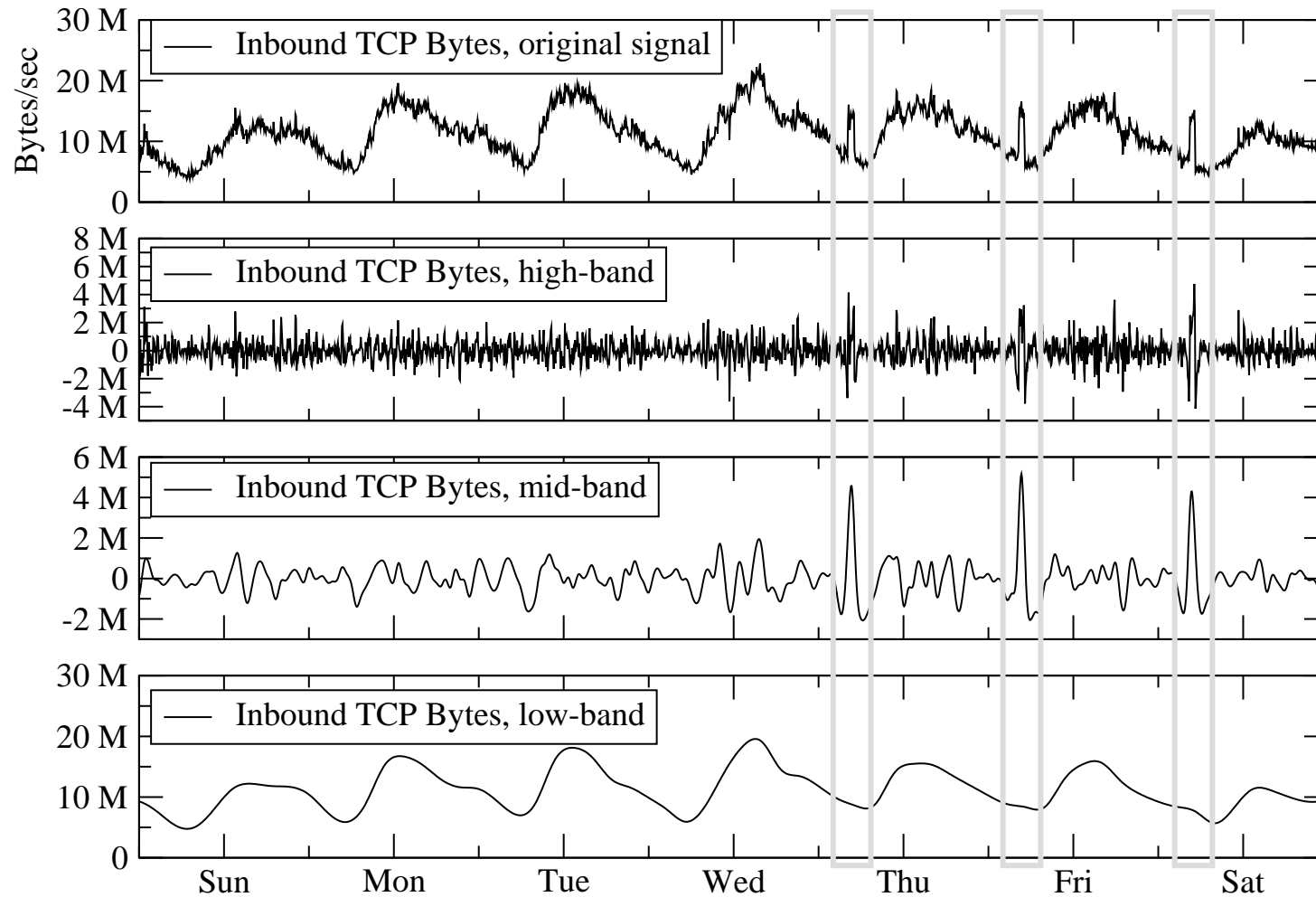
# Flow Traffic During DoS Attacks

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



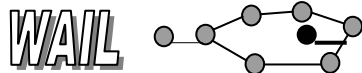
# Byte Traffic During Measurement Anomalies

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



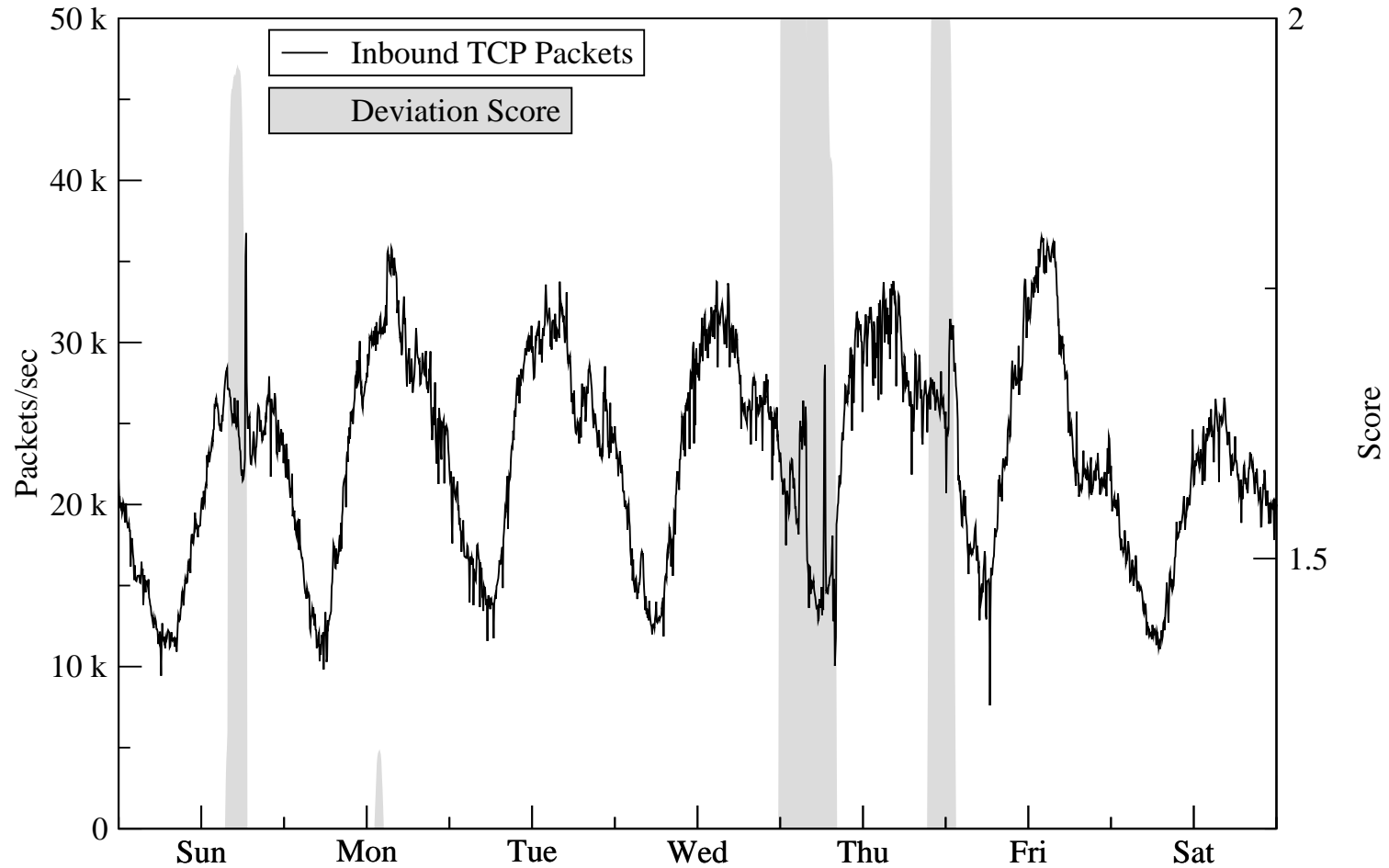
# Anomaly Detection via Deviation Score

- 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

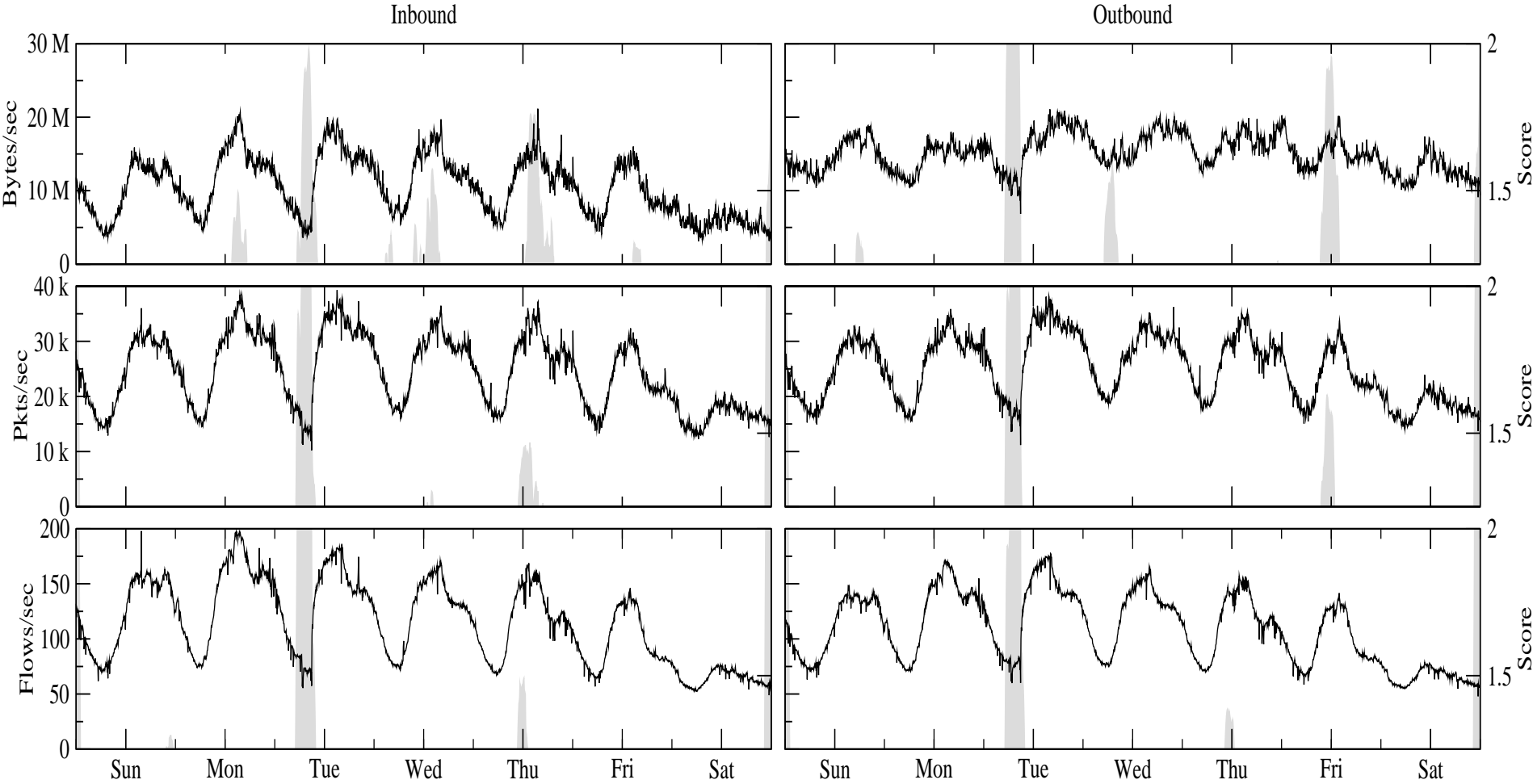


# Deviation Score for Three Anomalies

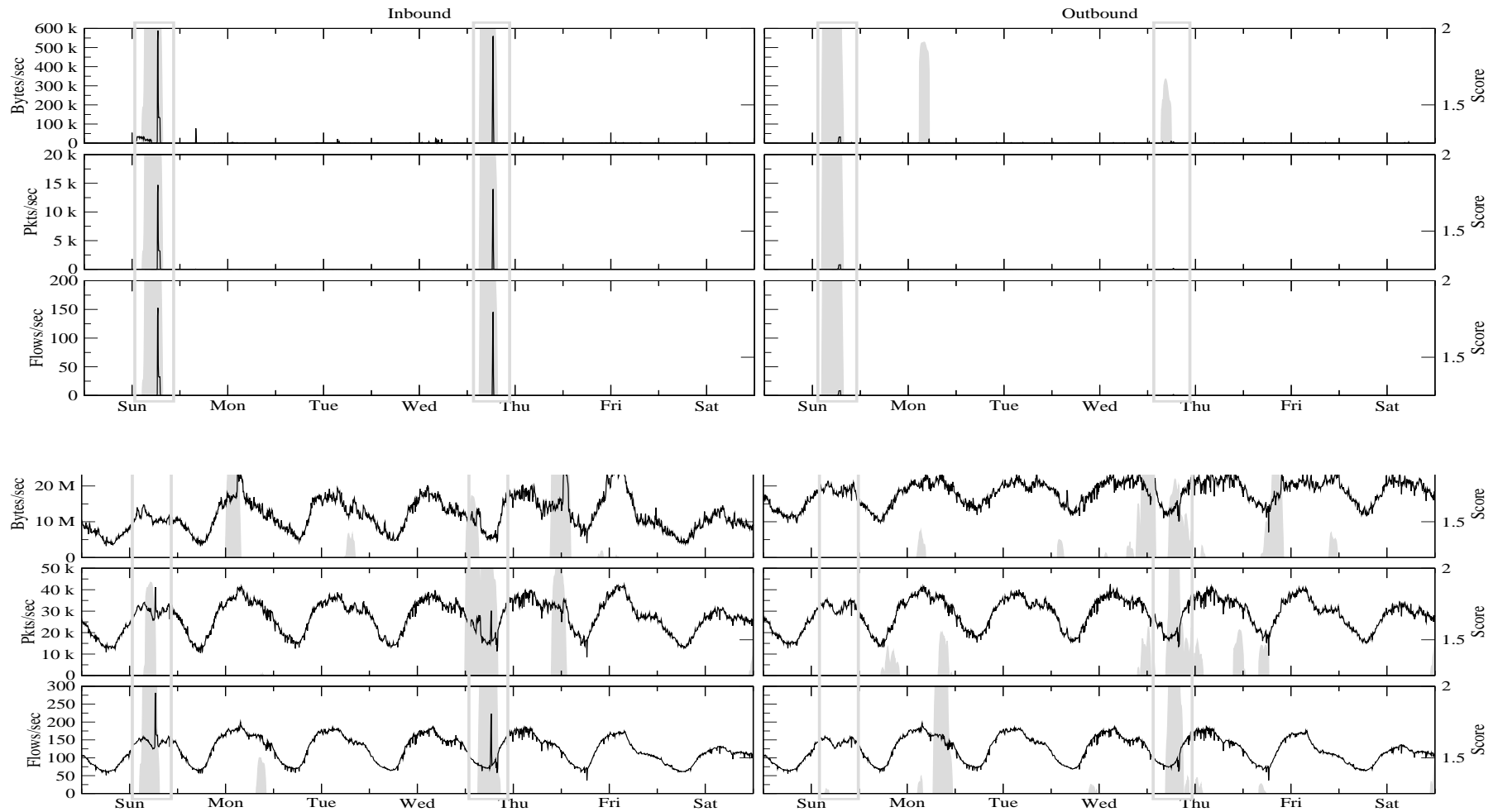
Campus TCP, Inbound, 2002-FEB-03 through 2002-FEB-10



# Deviation Score for Network Outage



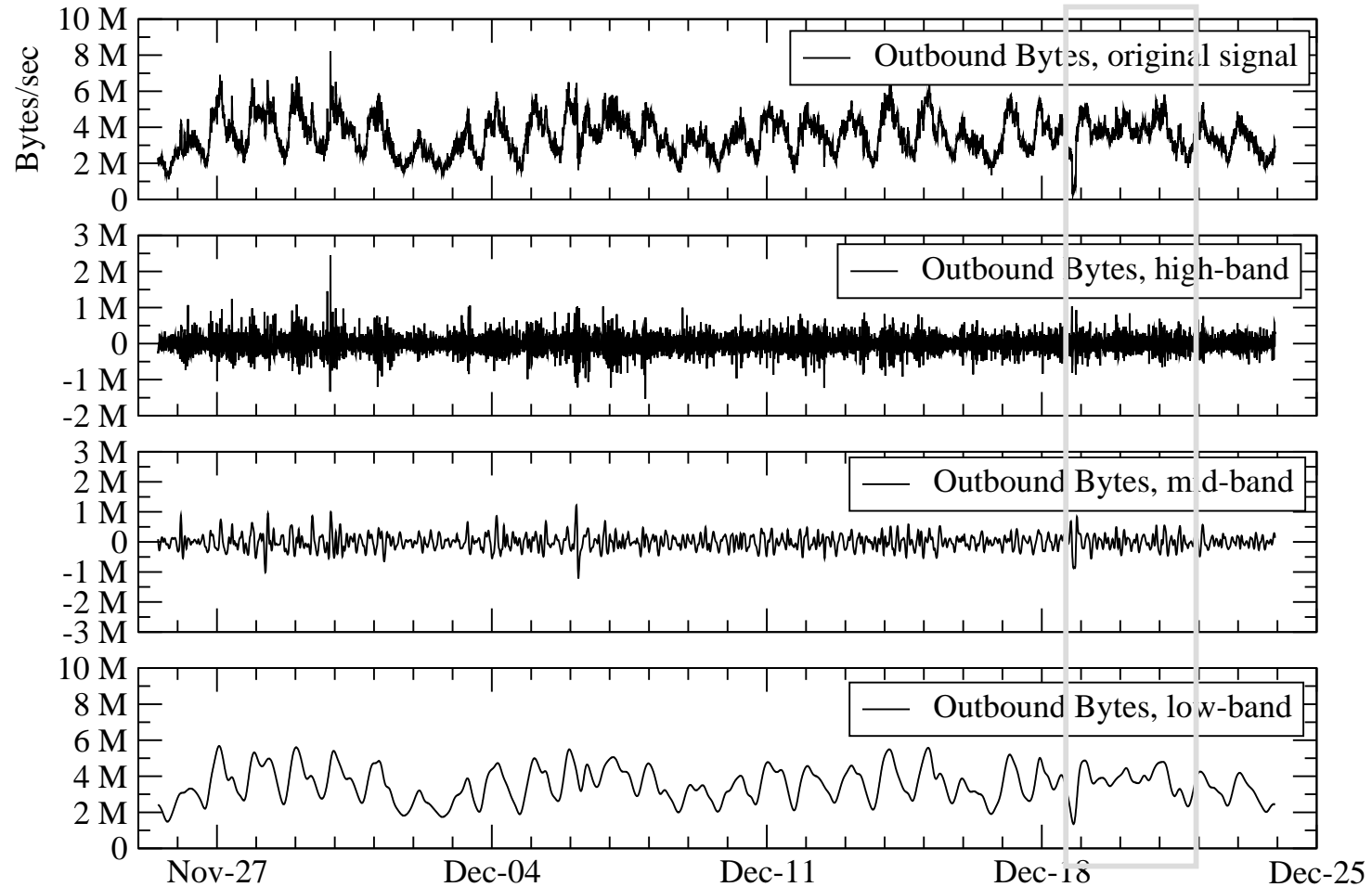
# Anomalies in Aggregate Signals





# Hidden Anomalies in Low Frequency

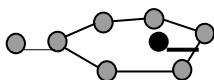
Class-B Network, Outbound, 2001-NOV-25 through 2001-DEC-23



# 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

