

SWINBURNE UNIVERSITY OF TECHNOLOGY

# Rapid Identification of BitTorrent Traffic

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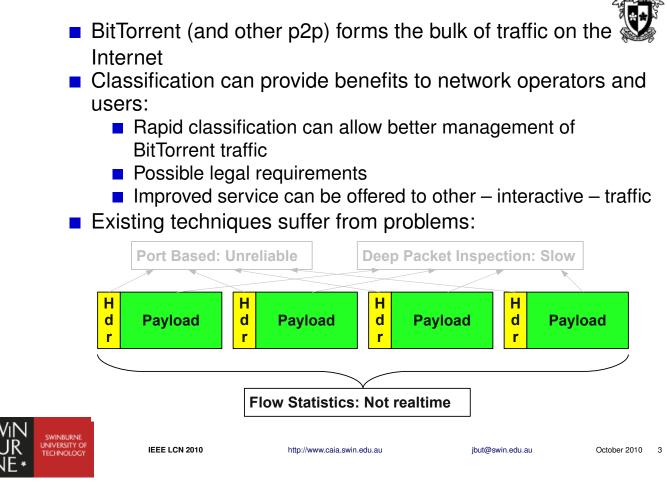
## **Outline**

- BitTorrent and Traffic Classification
- Traffic Observations
  - Statistical feature sets
- Classification
- Performance and classification timeliness

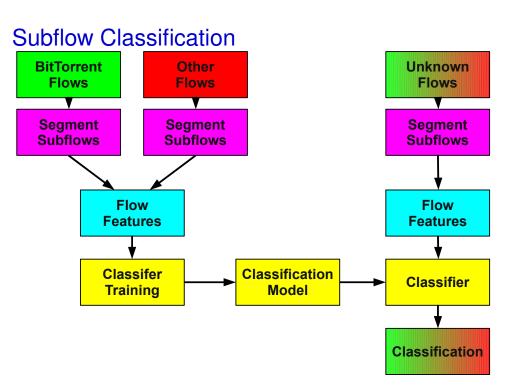




## **BitTorrent – Classification**



# Rapid Flow Classification



### We train the classifier to detect sub-flows





## **Common Packets**



## **Generic Packet Sizes**

- The BitTorrent protocol performs two functions
  - 1. Transfer data large packets
  - 2. Update status small packets
- BitTorrent exhibits packets of both types

#### **Data Flow**

- Centralised transfer applications are typically uni-directional
- For a p2p protocol, we expect traffic flow to be bi-directional

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# **Classification Features**



- A Characteristic BitTorrent Packet is one with a payload of 5 or 17 bytes
- *r*<sub>cbt</sub> = Ratio of Characteristic BitTorrent Packets to total packets within a (sub)flow

## r<sub>small</sub> – Small Packet Ratio

- A Small Packet is one with a payload of less than 40 bytes
- *r<sub>small</sub>* = Ratio of Small Packets to total packets within a (sub)flow



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## rlarge – Large Packet Ratio

- A Large Packet is one with a payload greater than 1350 bytes
- Based on Ethernet maximum segment size
- *r*<sub>large</sub> = Ratio of Large Packets to total packets within a (sub)flow

## $\sigma_{small}$ – Smaller Payload Standard Deviation

- Calculate the standard deviation of the TCP payload size for packets flowing in each direction
- If traffic flow is uni-directional, one of these values will be very small

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 $\sigma_{small} =$  Smaller of the two calculated standard deviations



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# Traffic Features

## **Traffic Traces**

- BitTorrent Captured from 4 swarms of up to 40 peers connected at ADSL1-like line rates, over 38,000,000 packets
  - Other University of Twente Public Traffic Trace<sup>1</sup>
    - FTP Captured distraction traffic nearly 1,000,000 packets from about 1,000 flows
- Our analysis indicates that these features differ
  Some features differentiate only two traffic types
- Features appear to differentiate BitTorrent (p2p) from other bulk transfer protocols (FTP)
- Features hold validity over sub-flows

<sup>1</sup>Available at: http://traces/simpleweb.org



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## Traffic Classification

- We calculated these four features over our data set for a www number of sub-flow sizes
- We used the WEKA implementation of C4.5<sup>2</sup> to train and test the classifier
  - Decision Tree based classification
  - Standard 10-fold Cross-Validation test
- Used captured FTP traffic as distractor traffic
- Results for whole-of-flow classification very good see paper for details
- Tested for all sub-flow sizes and each combination of features

<sup>2</sup>N. Williams, S. Zander and G. Armitage, "A Preliminary Performance Comparison of Five Machine Learning Algorithms for Practical IP Traffic Flow Classification", ACM SIGCOMM Computer Communication Review, vol. 36 no. 5 pp. 7–15, October 2006



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# Classifier Performance – individual features

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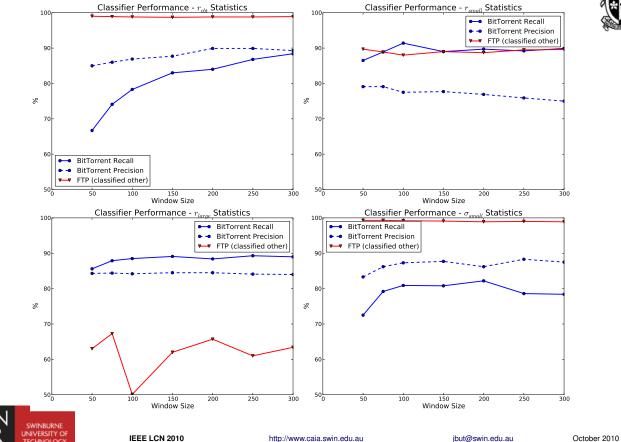


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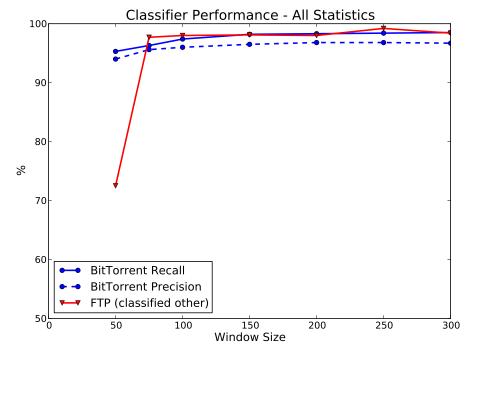


# Classifier Performance – all features



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# Classifier Performance – 150 packet subflows

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Classification Recall		
BitTorrent Recall	BitTorrent Precision	FTP (Recall as other)
98.2%	96.5%	98.1%

### Excluding *r*<sub>cbt</sub> feature

Classification Recall		
BitTorrent Recall	BitTorrent Precision	FTP (Recall as other)
97.5%	93.7%	97.6%

Minor drop in performance

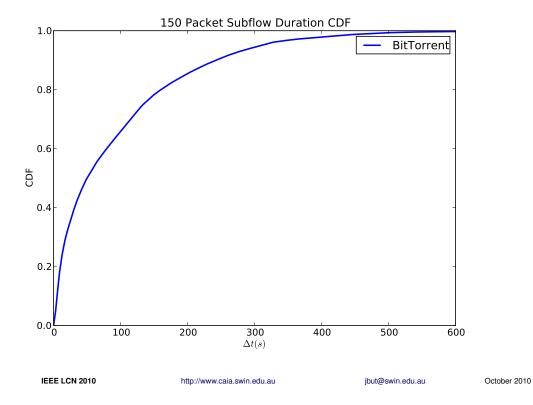
 Expected to be more robust for protocol changes or deliberate attempts to circumvent detection



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# **Classification Timeliness**

How long does it take to capture 150 packets for classification?





## **Conclusions**



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- Existing BitTorrent classification schemes are either non-scalable or use properties that preclude real-time classification
- We present four features  $r_{cbt}$ ,  $r_{small}$ ,  $r_{large}$  and  $\sigma_{small}$
- Suitable sub-flow features to allow for rapid classification
- Using an ML-based C4.5 classifier and these features:
  - Can classify entire BitTorrent flows with 98.9% Recall and 97.9% Precision
  - 150 packet sub-flows 98.2% Recall and 96.5% Precision
  - Ignoring  $r_{cbt}$  97.5% Recall and 93.7% Precision
- Analysis indicates that 150 packets subflows would typically be captured in under 3 minutes at ADSL like line rates
- Since these features are based on packet sizes, we expect this classifier to be robust for:
  - Encrypted BitTorrent
  - BitTorrent over UDP

