### CableMon: Improving the Reliability of Cable Broadband Networks via Proactive Network Maintenance

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### Broadband Networks are Important



#### **E-Commerce**



Work from Home



**Remote Learning** 



Entertainment

### Cable Broadband: One of Few Choices in U.S.

- Availability to U.S. homes
  - VDSL: 43%
  - FTTP: 29%
  - Cable: 93%













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- Aging can lead to problems
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- Reliablity of broadband is at most 99% (<< FCC's requirement for PSTN, 99.99%)

# Goal

#### Improving the reliability of cable broadband networks

### Roadmap

- Limitations of Existing Work
- CableMon Design
  - High-level Idea
  - Challenges
  - Solutions
  - ISP Deployment
- Evaluation
- Conclusion

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Cable Modem (CM)

Cable Modem Termination System (CMTS)

- Cable industry developed the PNM framework
- CMTS communicates with CM periodically to obtain PNM data
- A monitoring server collects all CMs' PNM data
  - E.g., Signal to Noise Ratio (SNR)



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## **Existing Work**

- Manually-set thresholds
  - A metric below/above a threshold: Repair the network
  - Hard to determine a proper threshold manually
  - Current recommendations: High false positives
    - In one of our studies, over 25% modems need repair following the PNM Best Practice document

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  - Save operational costs of ISPs by forecasting troubles before customer calls
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  - Efficient










### Tickets Correlate with Network Faults



# Challenges

- Tickets are noisy
  - Customers may call for network-irrelevant issues
  - Customers may not call when network faults occur

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  - Customers may call for network-irrelevant issues
  - Customers may not call when network faults occur
- PNM data
  - Instantaneous channel conditions, not sufficient for fault detection
  - Including environmental noise, not an accurate description of channel conditions

## Filter Non-network Related Tickets



- Tickets are filtered according to the Description, Action, Dispatched, etc.
- Network tickets correlate better with PNM data values















WMA Difference



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Variance



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- 1 ticket
- 10 data points (~40 hours)
- Ticketing Rate: 1/40 hours



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Ticketing Rate Ratio <u>Abnormal Ticketing Rate</u> Normal Ticketing Rate = 6.67





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Ticketing Rate Ratio = Abnormal ticketing rate Normal ticketing rate = 1.25



Argmax(Ticketing Rate Ratio)

Abnormal ticketing rate Normal ticketing rate

## **CableMon Properties**

- Ticketing Rate Ratio (TRR)
  - is monotonously decreasing w.r.t. both false positives (FPs) and false negatives (FNs) under the assumption ticketing noise is uniformly distributed
  - 2. is maximized iff. both FPs and FNs are 0

Formal proof can be found in the paper

# Selecting Top Features

- We select the features that achieve the highest ticketing rate ratio
- For the features generated from the same metric, we will select at most two

### **Top Five Features**

Feature	<b>Ticketing Rate Ratio</b>
snr-var	14.49
uncorrected-var	7.66
rxpower-wma-diff	5.31
t3timeouts-wma-diff	4.93
t4timeouts-var	4.18

## **Combining Features**

- If one feature is abnormal, the data point is labelled as abnormal
  - Different features may detect different types of network faults
  - Use the threshold(s) that maximize(s) the ticketing rate ratio for each feature

## How ISPs Use CableMon

- Proactive detection
  - Make a dispatch decision only after a fault persists
  - Convert pointwise detection to abnormal event detection using a sliding window algorithm
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- Diagnosis
  - Determine whether the customer reported problem is network relevant

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### Evaluation

• Dataset



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• Other Approaches
• Dataset



• Other Approaches



**ISP** Tools

• Dataset



• Other Approaches



• Dataset



• Other Approaches



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• Other Approaches



- Methodology
  - Ideal: Real world deployment (working in progress)
  - Experiments: Emulate how ISPs use CableMon
    - Run the sliding window algorithm
    - Count the number of tickets arrived during an abnormal event
- Metrics
  - Ticket prediction accuracy
  - Ticket coverage
  - Normalized ticketing rate

#### Results

	Ticket Prediction Accuracy	Ticket Coverage	Normalized Ticketing Rate
CableMon	81.92%	22.99%	3.55
SVM	75.64%	12.54%	2.02
Random Forest	73.14%	14.21%	2.24
Decision Tree	68.93%	15.53%	2.52
Comcast's Tool	23.48%	2.21%	1.18
AnonISP's Tool	10.04%	25.13%	0.98

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#### More results can be found in the paper

### Conclusion

- CableMon to detect network faults
- Use tickets as hints: No manual labeling
- Overcome the noise from both PNM data and customer trouble tickets
- Achieve high ticket prediction accuracy, and moderate ticket coverage

# Thanks for your attention!

# **Questions?**