# Towards application of network topology information to network causal log analysis 

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## Difficulty of leveraging system log

- Huge dataset
- Large scale and complicated systems
- 150,000 lines / day in SINET 5
- Automated analysis required

- Difficult to analyze automatically
- Free-format
- Mixture of frequent and sparse logs
- Lengthy / Repeated data


## Automated analysis of system log

3 challenges for operating network


- Anomaly detection
- State modeling [1,2]
- Bayesian estimation [3]

[1] K. Yamanishi et al. "Dynamic syslog mining for network failure monitoring". In ACM KDD’05, p. 499, 2005.
[2] F. Salfner et al. "Using hidden semi-Markov models for effective online failure prediction". In IEEE SRDS, pp. 161-174, 2007.
[3] P. Chen et al. "Causeinfer: Automatic and distributed performance diagnosis with hierarchical causality graph in large distributed systems". In IEEE INFOCOM, pp. 1887-1895, 2014.
[4] I. Beschastnikh, et al. "Inferring Models of Concurrent Systems from Logs of Their Behavior with CSight." In ICSE 2014, 468-479, 2014.
[5] T. Kimura et al. "Spatio-temporal factorization of log data for understanding network events". In IEEE INFOCOM, pp. 610-618, 2014.


## Causal analysis of system log

- Graph-based causal inference ${ }_{\text {[9] }}$
- PC algorithm: causal structure estimation
- Exploratory analysis in contrast to existing approaches [7, 8]
- Challenges
- Processing time
- Reliability of detected information
> Improve causal analysis based on topology knowledge
[6] B. Tak et al. "LOGAN: Problem Diagnosis in the Cloud Using Log-Based Reference Models," in IEEE IC2E, 2016, pp. 62-67.
[7] Z. Zheng et al. "3-Dimensional root cause diagnosis via co-analysis," in ACM ICAC, 2012, pp. 181.
[8] A. Mahimkar et al. "Towards automated performance diagnosis in a large iptv network," in ACM SIGCOMM, 2009, pp. 231-242.
[9] S. Kobayashi et al. "Mining causality of network events in log data", IEEE TNSM, vol. 15, no.1, pp. 37-67, 2018.


## Goal

- Extract causality of events in system logs efficiently
- Based on causal inference (PC algorithm)
- Using network topology knowledge
- Provide reliable information for system management and troubleshooting
- More accurate information
- Less redundant (or meaningless) information


## Dataset

- SINET4
- https://www.sinet.ad.jp/en/top-en
- A nation-wide R\&E network in Japan
- 8 core routers and 100 over L2 switches
- 15 months syslog data
- 3.5 million lines to analyze


## Causal Inference

- Conditional Independence
- A and B are independent if the effect of confounder $C$ is excluded
- $A$ and $B$ are conditionally independent given $C$
- PC algorithm [10]
- Directed acyclic graph (DAG)

- Explore conditional independence and remove false edges
[10] P. Spirtes et al. "An algorithm for fast recovery of sparse causal graphs", Social science computer review, vol. 9, pp. 62-72, 1991.


## Flow of PC algorithm

Complete graph (initial)


Skeleton graph


Directed acyclic graph


- Remove edges of conditional independence
- Statistical test for conditional independence [11] s) [12]
- G2 test (for binary or multi-level data)
- Fisher-Z test (for continuous data)
[11] R. E. Neapolitan. "Learning Bayesian Networks." Prentice Hall Upper Saddle River, 2004.
[12] T. Verma, et al. "An algorithm for deciding if a set of observed independencies has a causal explanation". In Proceedings of UAI'92, pp. 323-330, 1992.


## Log analysis and causal inference ${ }_{\text {[9] }}$

Oct 17 17:00:00 routerA System shutdown by root
Oct 17 17:00:05 switchB Error detected on eth0
Oct 17 17:00:15 routerC BGP state changed from Established to Idle
Oct 17 17:00:15 routerD SNMP trap sent to routerA

[9] S. Kobayashi et al. "Mining causality of network events in log data", IEEE TNSM, vol. 15, no.1, pp. 37-67, 2018.

## System architecture



## System architecture



## Proposed method

- Preprocessing based on network topology
- Heuristic: Only network events of connected devices have causal relations
- Edit initial graph of PC algorithm
- Complete graph -> Pruned graph

Complete graph (initial)


Skeleton graph

Directed acyclic graph


## Proposed method

- Preprocessing based on network topology
- Heuristic: Only network events of connected devices have causal relations
- Edit initial graph of PC algorithm
- Complete graph -> Pruned graph

Pruned graph (initial)


Skeleton graph


Directed acyclic graph



## Generate analytics topology

- Consider both L2 and L3 connectivity
- Add edges of L3 connections



## Pruning initial graph

- Prune edges between events of unconnected devices in given analytics topology

Complete graph of events
Topology knowledge


New initial graph for PC algorithm

## Compare with existing method

- Existing method: Area-based separation ${ }^{\text {(9] }}$
- Multiple smaller complete graphs

Area-based


Topology-based


- Edges among core routers
- Sparse edges


## Evaluation

- Generate causal DAGs for 455-days log data
- 35 million lines
- 1789 log templates, 132 devices
- Results

| Method | Edges | Time <br> (sec/day) |
| :---: | ---: | ---: |
| None | 30,174 | 1,220 |
| Area-based | 29,195 | 870 |
| Topology-based | 26,005 | 940 |

Both methods decrease $25 \%$ processing time

## Classification of causal edges

| Type | Nodes | Ends of edges |  |  |
| :---: | :---: | :---: | :---: | :---: |
|  |  | None | Area | Topology |
| System | 53,505 | - 1 co | O-0n7 |  |
| Network | 12,901 | a monitoring server |  |  |
| Interface | 13,446 |  |  |  |
| Service | 7,697 | 742 | 435 | \07 |
| Mgmt | 75,677 | 25,183 | 23,722 | 17,359 |
| Monitor | 2,452 | 267 | 305 | 298 |
| VPN | 3,465 | 50 | 1,074 | 106 |
| Rt-EGP | 3,831 | 1,576 | 1.8 | 1,605 |
| Rt-IGP | VPN connections in core routers |  |  |  |
| Total | -> Condition | indepen rea-based | dence no method | effective in |

## Case study 1

- Found causal edges in topology-based method (and NOT in area-based method)
- Events of system errors


Removed in area-based method because of conditional independence of failed (impossible) confounding factor

## Case study 2

- Found causal edges in topology-based method (and NOT in area-based method)
- Events of VPN among core nodes

```
1771[vpn] in core router B
**:rpd[**]:
RPD_MPLS_LSP_SWITCH:
MPLS LSP ** switch
from secondary(**) to
primary(**), Route ...
```

```
1771[vpn] in core router C
**:rpd[**]:
RPD_MPLS_LSP_SWITCH:
MPLS LSP ** switch
from secondary(**) to
primary(**), Route ...
```

Topology-based method can find edges between core-router events (and Area-based method cannot)

## Discussion

- Depends on network topology
- Not effective in SINET5 (Full mesh topology)



## Concluding remarks

- Estimate causal relations among network events in SINET4 log data
- Use topology knowledge of network devices to prune initial edges of PC algorithm
- Decrease 25\% processing time
- More accurate analysis than area-based method
- Future work
- Co-operative analysis with other data sources
- Layer-based preprocessing

