Automating Network Application Dependency Discovery: Experiences, Limitations, and New Solutions

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Enterprise network management is complicated



- 1,000s network applications
- 1,000s staffs in IT support
- \$\$ millions of dollars spent every year



Why enterprise network management is complicated



Challenges in discovering service dependency

Heterogeneous applications

- Functionality
- Deployment setups
- Knowledge distributed across layers and locations
- Applications evolve continuously



Why service dependency is useful



Current solutions

- Based on human knowledge
 - Expensive
 - Error-prone
 - Hard to keep up-to-date information
- We need automated solution for dependency extraction



Related work

- Co-occurrence based dependency discovery
 Sherlock & eXpose [SIGCOMM'07 & 08]
- Execution causality path extraction
 Project 5 [SOSP'03] & WAP5 [WWW'06]



Our contributions

- Introduce a new technique to discover dependencies based on *spike in delay distribution*
- Identify the limitations of dependency discovery based on temporal analysis
- Evaluate our technique on five dominant applications in Microsoft's enterprise network
- Significantly improve the accuracy of dependency discovery over prior work



Outline

Overview

- Dependency discovery techniques
- Deployment & results
- Conclusion



Design goals

- Applicable to variety of applications
 - Passive sniffing
 - Only parse into TCP/IP headers
- Minimizing false dependencies without losing true dependencies
 - Hard to recover missing true dependencies
 - Minimize the effort to filter false dependencies



Orion

 Key idea: time delay between dependent services reflects typical processing and network delay





Dependency discovery process





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Inferring application messages

- Problem: dependency exists between application messages
- Only rely on TCP/IP headers
 - Aggregate packets into flows
- Benefits
 - Reduce bias introduced by long flows
 - Reduce number of pairs



Dealing with scalability

- Service: (ip, port, proto)
- Problem:
 - Too many service pairs
- Solutions
 - Ignore transient "services"
 - Only consider service pairs that are close in time



Filtering noise in delay distributions





Overcoming a lack of samples

- Problem: Orion requires fair number of samples to infer dependency
- Solution:
 - Client aggregation
 - clients have similar dependencies
 - Server aggregation
 - Same application hosted on a cluster of servers
 - Port aggregation
 - Same service hosted on different ports



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Orion deployment



- Five dominant applications
 - Exchange, Office Communicator, Source Dept, Distributed File System, Web
- Traffic on R0
 - Over 2,000 clients
- Traffic on R1/R2
 - Email service used by over 10,000 users
 - Largest internal web portal



Accuracy of dependency discovery

- Evaluation criteria
 - Missed dependencies false negative
 - Incorrectly-inferred dependencies false positive
 - Reduction ratio 97.9% 99.6% for Orion

	Exchange		DFS		Sharepoint		OC client		SD client	
	FN	FP	FN	FP	FN	FP	FN	FP	FN	FP
Orion	0	26	0	13	0	3	0	77	0	4
Sherlock:10	0	178	0	102	0	65	2	125	0	52
Sherlock:100	0	57	0	93	0	168	1	85	0	29
eXpose	1	443	0	570	0	565	1	1416	0	323



Dependencies of Office Communicator



OC client dependencies



Dependencies of Exchange



Exchange client dependencies



Effects of noise filtering & flow generation

	Exchange		DFS		Sharepoint		OC client		SD client	
	FN	FP	FN	FP	FN	FP	FN	FP	FN	FP
Orion	0	26	0	13	0	3	0	77	0	4
noFilter	0	49	0	25	0	6	0	159	1	19
noFlow	0	2488	0	988	0	534	0	3594	0	198



Convergence





Impact of aggregation

DFS: distributed file system



Number of Clients



Limitations

- Relatively long discovery time
- Not applicable to P2P applications
- May miss certain types of interactions
- May include false positives



Conclusion

Lessons learned

- Temporal analysis has inherent limitations
- False positive can be reduced to a manageable level

Summary

- A new technique to discovery dependency based on spike in delay distribution
- Evaluate using production enterprise applications



Thank you!

Questions?



