

Risk in Cyber Systems



Marshall Kuypers

PhD Candidate, Department of Management Science and Engineering
Stanford University
mkuypers@stanford.edu

Dr. Elisabeth Paté-Cornell

Professor, Department of Management Science and Engineering
Stanford University
mep@stanford.edu

Notes meant for voice track are in blue bubbles

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Our research is motivated by the idea that...

significant uncertainty surrounds cyber security investments

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An organization considering three investments currently does not have a rigorous way to assess the value of different safeguards, or to quantify cyber risk.













Two-factor authentication

Subscription for threat intel

Data loss prevention

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Likelihood

Organizations use 'people sitting around a table' to make decisions, or rely on handwavy explanations from security vendors.

Current methods are limiting

PSAT



Hand Waving



Medium High **Extreme** Very likely Medium High Low Likely Low Low Medium Unlikely 2 What is Minor Moderate Major the chance it will happen?

Impact

The cybersecurity framework in action: an Intel use case

			ENDPOINT/ DATA			
IDENTIFY	POLICY	NETWORK	PROTECTION	IDENTITY	OPs	
Business Environment	3	3	3	2	~	
Asset Management	3	2	2	2 (1	
Governance	3	2	3	2 ,		
Risk Assessment	~	2	2	2	2	
Risk Management Strategy	4	3	2	2 /	2	
PROTECT						
If a method exists, it is like	ly	3	3	2 2	3	
to be qualitative: Intel		Mapping highlighted outlier and major differences				
published an example, but		2	1	\ <u>`</u>	1	
the analysis may not have						
been data driven						



rigorous, quantitative methods now exist

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Quantitative approaches lead to more insights

Data Analysis

Significant data exists in organizations!



Malicious Insider



375

Website Compromises

Modeling

Use dollars



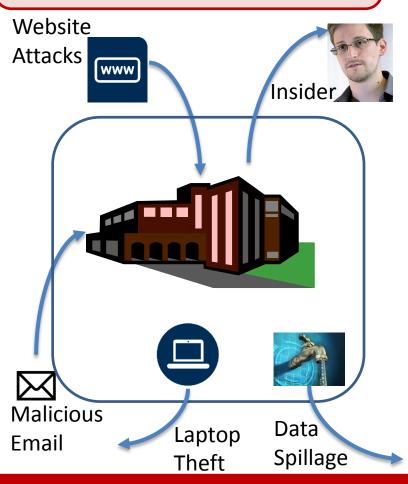




Use distributions, not averages

Our method is data driven, uses dollars, and uses distributions. Overall, we model the frequency and impact of different cyber attack categories and quantify risk

Risk Analysis





Our work has been successful in part because we've gotten access to security incident data. These...

incident databases are treasure troves of intel

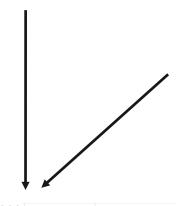
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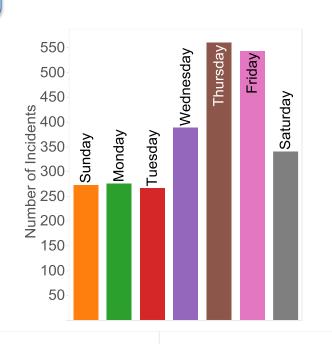
We can analyze shellshock attacks

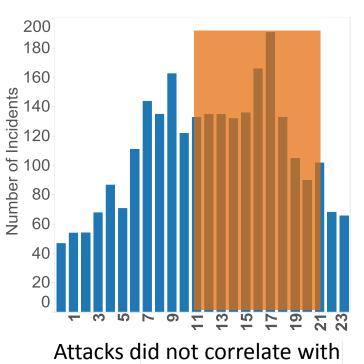
Shellshock attacks

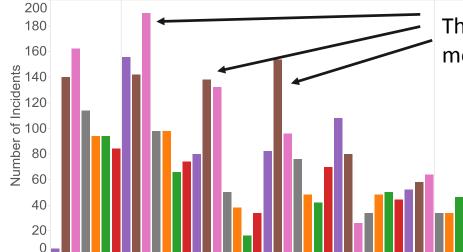
Shellshock publically announced on September 24th



Within 5 hours, a shellshock attack was detected







Thursday and Friday were the most common days for attacks

US workday hours

Incidents continued to occur for several months



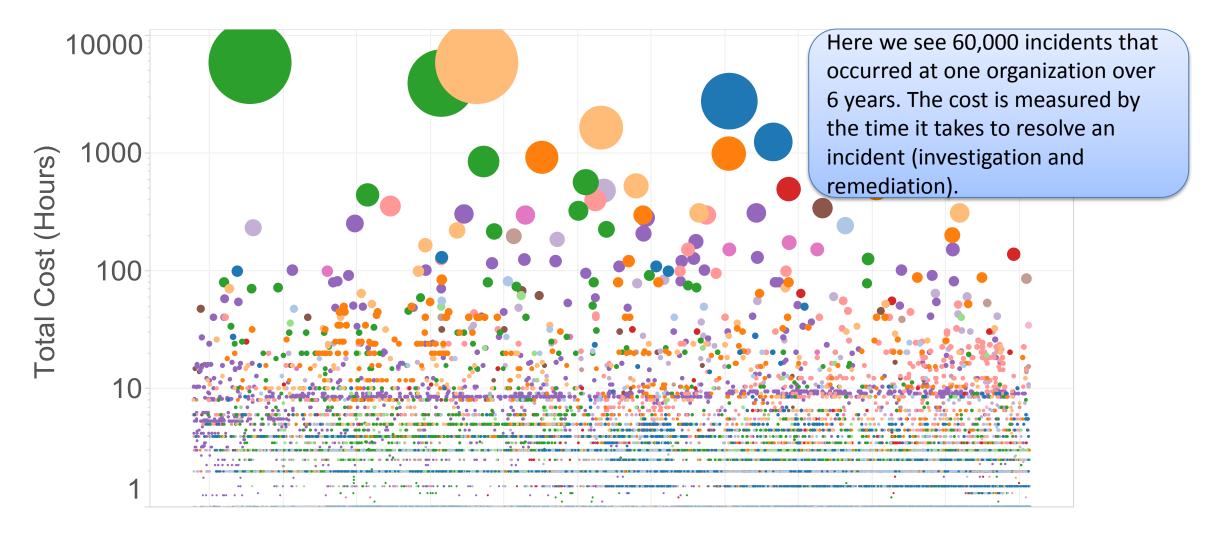
We can also do a really good job of quantifying the frequency and impact of cyber security incidents.

frequency and impact of cyber incidents can be quantified

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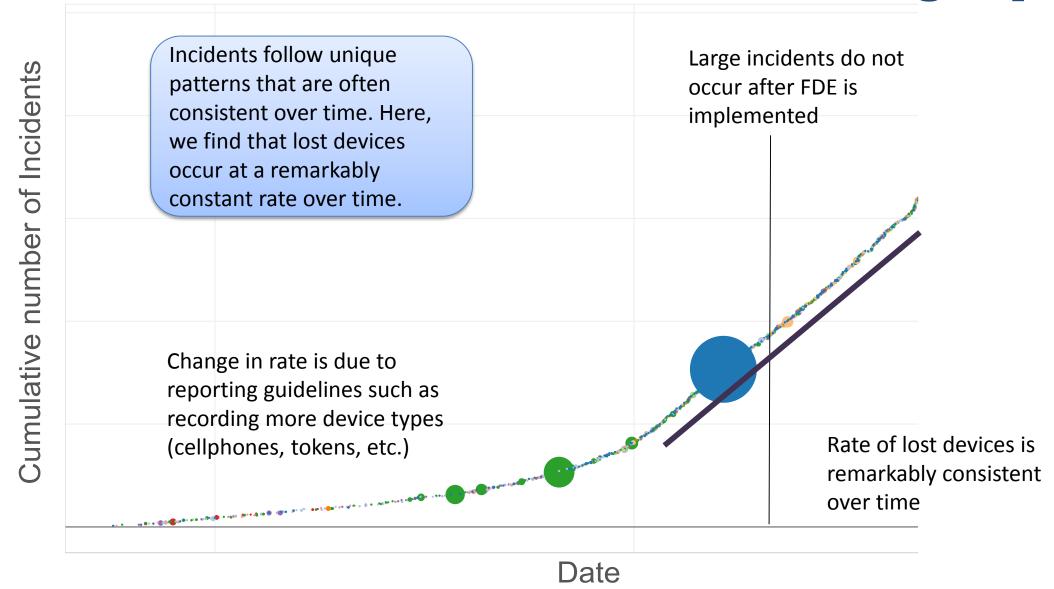
Most incidents take less than 100 hours to resolve



Date



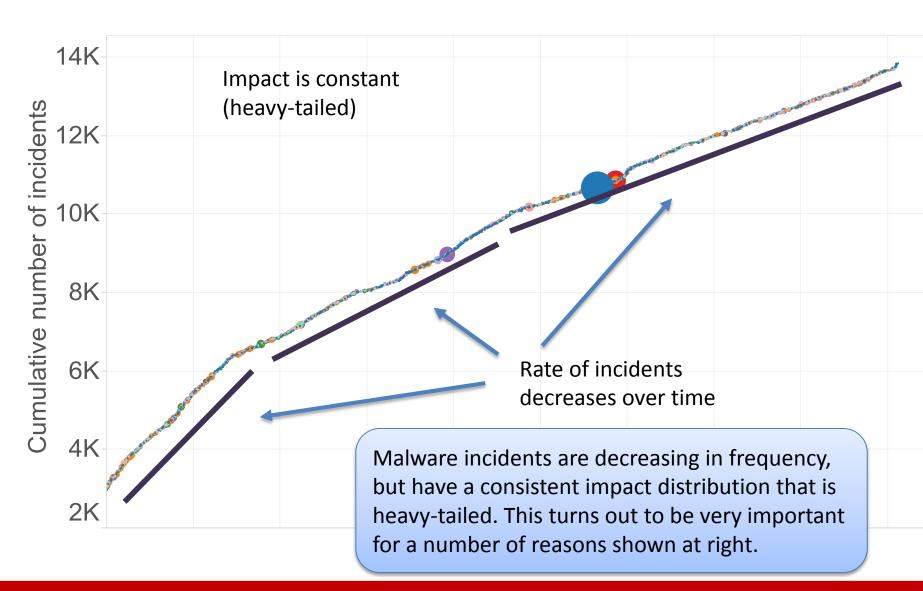
Lost devices: constant rate, decreasing impact



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Malware: Decreasing rate, constant impact



Large Events are NOT outliers

No 'average' or 'typical' cyber breach

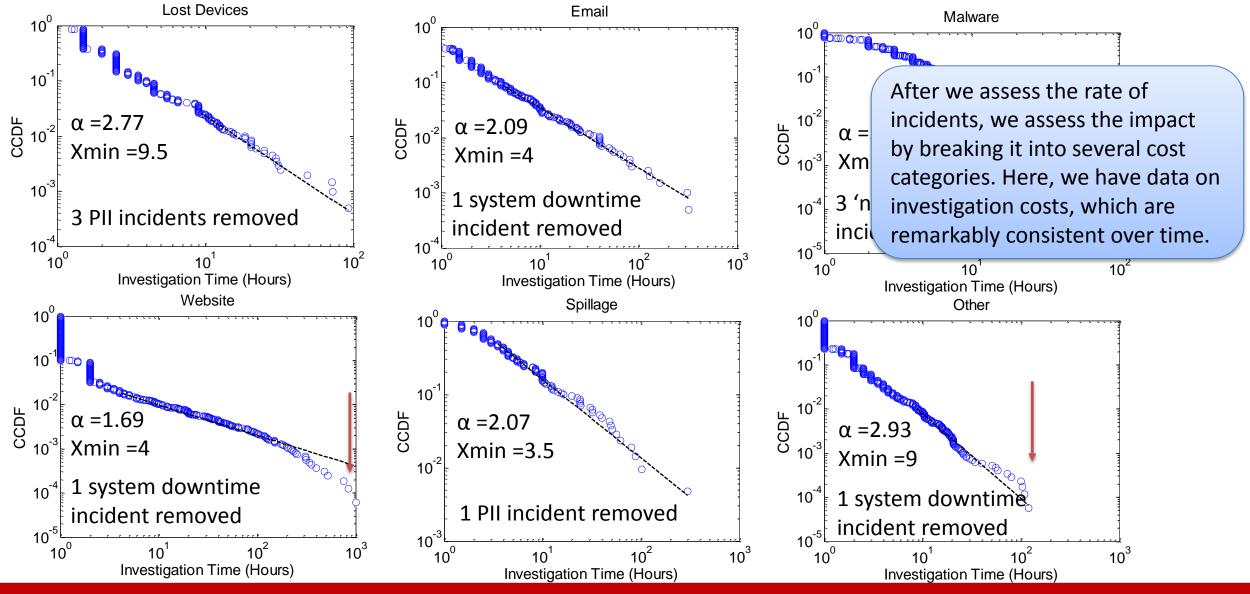
Standard deviations and some risk metrics (value at risk) are not valid



Largest incident can be more impactful than all other incidents combined!



Investigation is a major cost, and can be quantified





Reputation damage uncertainty is modeled



Mars Global Surveyor

Failure: 2006

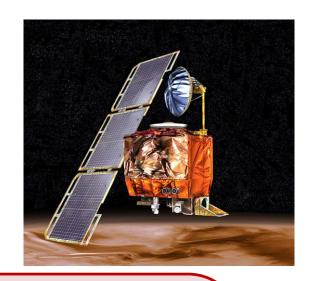
Cost:\$154M to build, \$65 to launch, \$20M per year to

operate

Description: Software update

error causes computer crash and

fried batteries



Mars Climate Orbiter

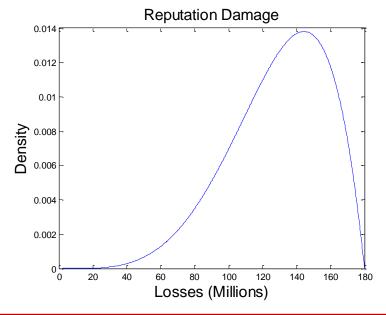
Failure: 1999 Cost:\$193M

Description: metric and standard units conversion crashes the orbiter into

mars



Reputation damage has been a hurdle in the past, but we explicitly model the uncertainty of losses (seen at right). For a case study, take chip manufacturer that stocks satellite parts. We can look at failures of satellites (that are cyber attack flavored, not attacks) to estimate costs. Academic research shows that stock prices only fall for 2 days after a breach, and we can look at Target, RSA, or Sony for other case studies.





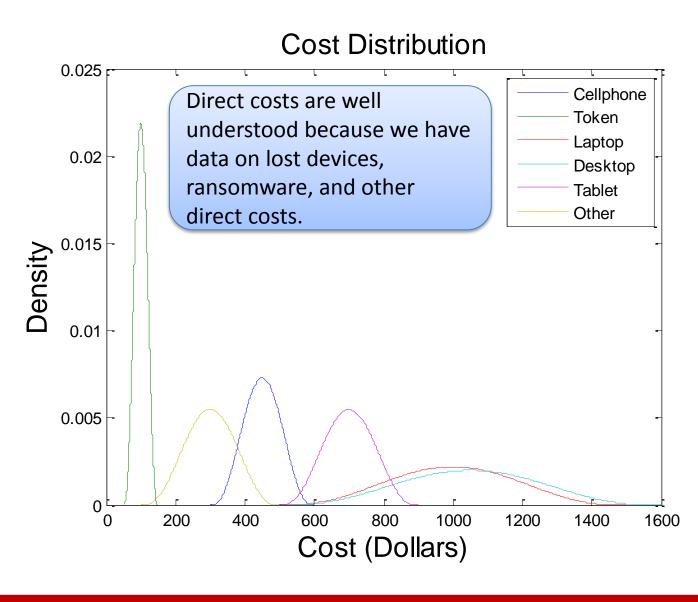
Direct costs are well understood

Probability	Device	Average Cost
0.34	Cellphone	\$400
0.32	Token	\$100
0.20	Laptop	\$1000
0.07	Other	\$300
0.05	Desktop	\$1000
0.02	Tablet	\$700



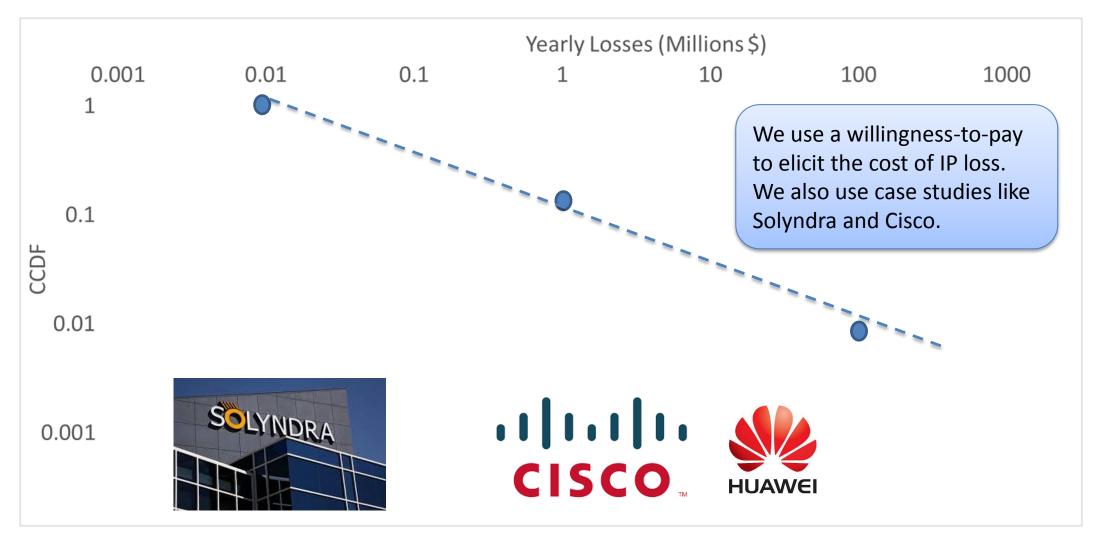
Equipment Losses







Willingness-to-pay used for intellectual property losses





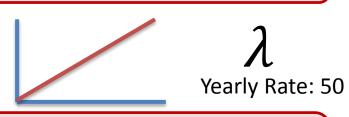
rolling this **information** together, we can obtain excellent **risk assessments**

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A case study demonstrates the method

Rate of spillage incidents



Impact Distributions (Data Spillage)

Investigation

Alpha 1.22, scale 0.827

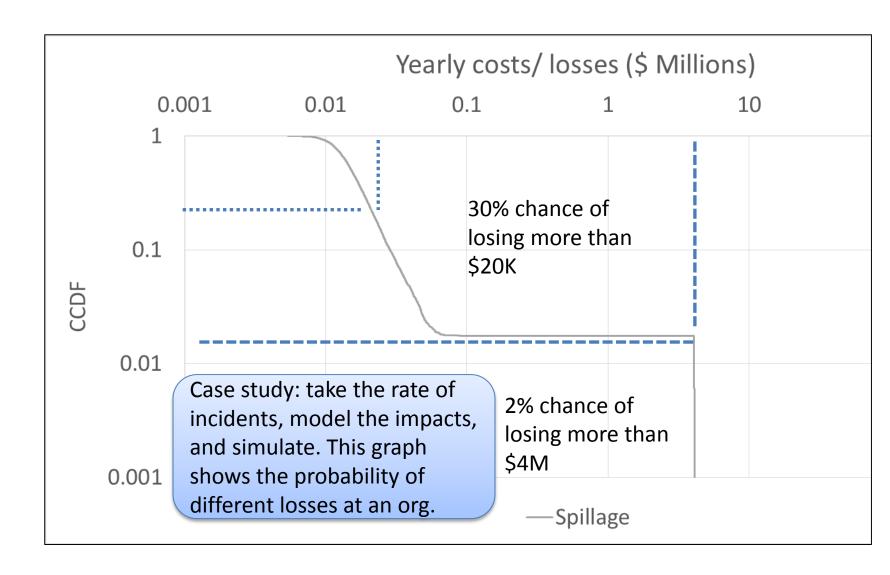
Reputation

Distribution

Fines

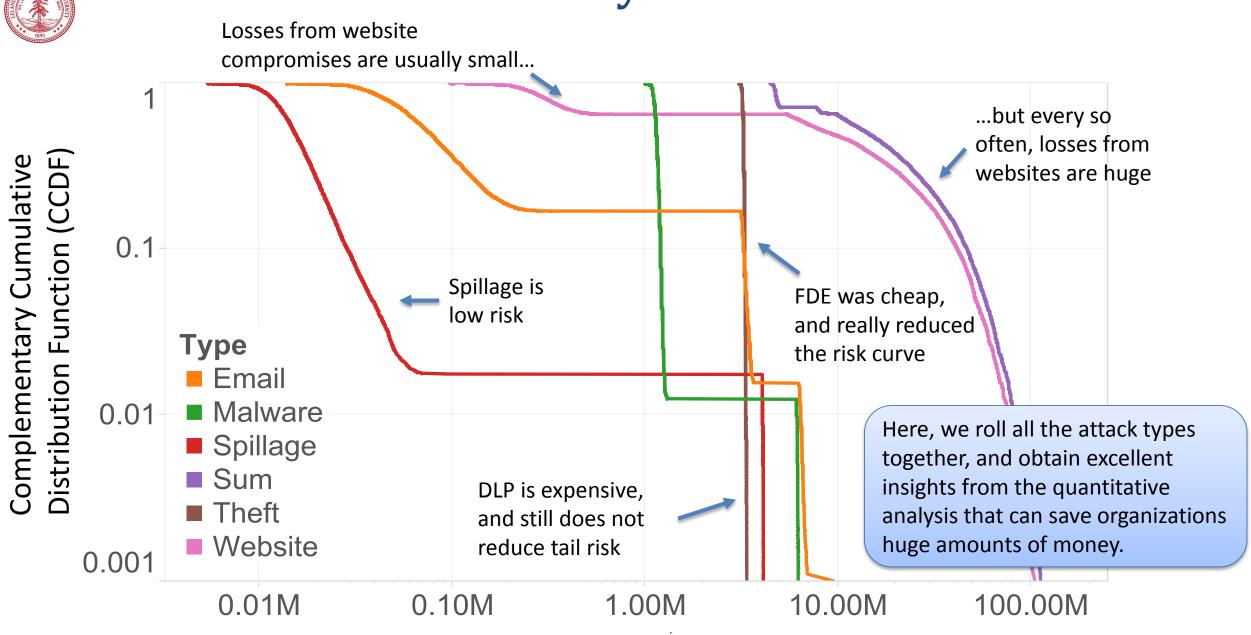
IP Loss







A case study demonstrates the method



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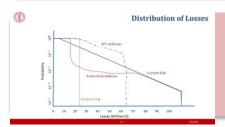
Conclusions



Probabilistic risk analysis methods inform **actionable** decisions.



Incident data is priceless.



Safeguards can be compared and prioritized.



Monetary impacts help justify **budgets** and **communicate** risk.