

## STIX Analytics-- From Threat Information Sharing to Automated Response

Secure and Resilient Cyber Ecosystem Industry Workshop Presentation for DHS

Dr. Ehab Al-Shaer (PI), Dr. Bill Chu (PI) University of North Carolina Charlotte

Ealshaer, billchu@{uncc.edu,ccaa-crc.org}



#### **Agenda Outline**

- Motivation of Cyber Threat Information (CTI) Sharing
- Background STIX (if needed)
- Challenges to Effective CTI Sharing
- Our Research Proejcts/Directions
  - STIXChecker
    - Logical Formalization of STIX (OWL/SMT) and configurations
    - Impact analysis
  - STIXAnalytics
    - Determining Network Relevance
    - Visual Analytics
    - Reputation Analysis
  - ThreatMitigation
    - Impact Analysis: Killing the Cyber Kill-Chain
    - From STIX to Actions

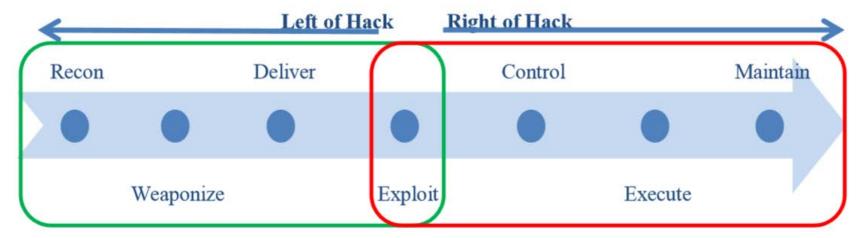


#### What is STIX

- A language to specify, capture, characterize and communicate
   Cyber Threat Information.
- A standardized and structured way to represent threat information
- Both human readable and machine parsable.
- Built upon active participation and feedback from a broad spectrum of organizations and experts linked with government, academia and industry.
- Initial implementation has been done in XML Schema and JSON.
  - Plan to iterate and refine with real-world use



#### **STIX Concept**



**Example Cyber Kill Chain** 

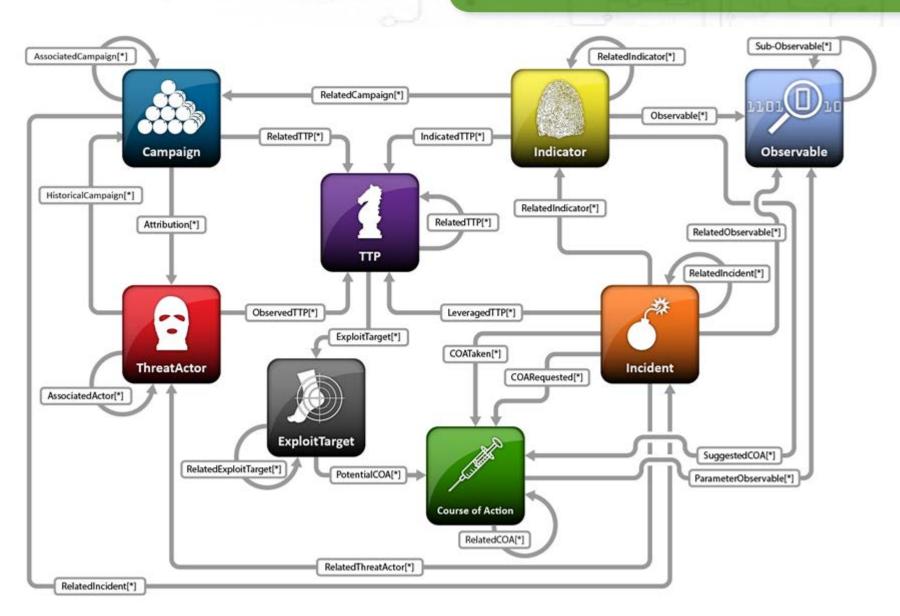
- Proactive approach to resist adversary, preferably before the exploit stage.
- Only possible through adoption of Cyber Threat Intelligence



#### STIX Embedded with CTI

#### Consider these questions: What activity are we seeing? What threats should I look for on my networks and systems and why? Where has this threat been seen? What does it do?-What weaknesses does this threat exploit? ExploitTarget Why does it do this?-Who is responsible for this threat?-What can I do about it? -RSACONFERENCE 2013 @ 2013 The MITRE Corporation. All rights reserved







# CHALLENGES TO EFFECTIVE CTI SHARING



#### Challenges

- ❖ Intelligence must be actionable otherwise it is useless. [CTI rules]
- ❖ Making Threat Intelligence Actionable. [RSA Conference 2015]
- ❖ Stix XML leads to interpretability and portability issues.
- ❖ Difficult to import as it is in existing analysis tools.
- Implementation independent solution is highly favourable.
- Automated inference and reasoning deficiency in XML.





#### Challenges

- For the Network Admin (use-case 1)
  - STIX feeds requires extensive analysis to extract elements relevant to the network.
  - Mapping threats to their counter measures is a manual process and lacks cost-benefit and impact analysis.
- For the Cyber-Security Analyst (use-case 2)
  - Visualization of the 'big picture' of the cyberthreats landscape



#### **Challenges**

- Identified Problems when Stix mapped or used for a particular network.
- Thousands of threats shared every day using Stix.
  - How to identify threats relevant to organization infrastructure?
  - ❖Which one is important?
  - ❖Which has higher impact or critical ?
  - ❖What is the likelihood of particular exploit ?
  - ❖ What could be the damage in terms of privacy, integrity, availability.
  - How much Cost will be affected?
  - What nodes will be affected, if particular threat occurs.



# STIXCHECKER – FROM STIX TO ACTION



#### STIXChecker Objectives

- Extend and develop ontologies as a working model for STIX, network and vulnerabilities
- Identify relevance of prevalent STIX threats according to network architecture.
- Quantitative estimation of the **impact** induced by STIX threats to the enterprise mission, assets and security requirements.
- Automatic transition from CTI to mitigation actions.
- Cost-benefit mitigation analysis to achieve an optimal level of security when provided with a limited budget.



#### **STIXAnalytics Objectives**

#### Risk Analytics:

What is the impact of STIX-threats on the enterprise policy based on its network configuration and vulnerability scanning reports?

• Intelligence-Driven Proactive Cyber Defense:

What are the configuration changes and vulnerability fixes that will reduce the risk to an acceptable level without affecting the mission of the system?

#### Visual Analytics

What are the most prevalent threats? How are they related? Which ones are instances of the same attack? Which bots co-host multiple malwares?

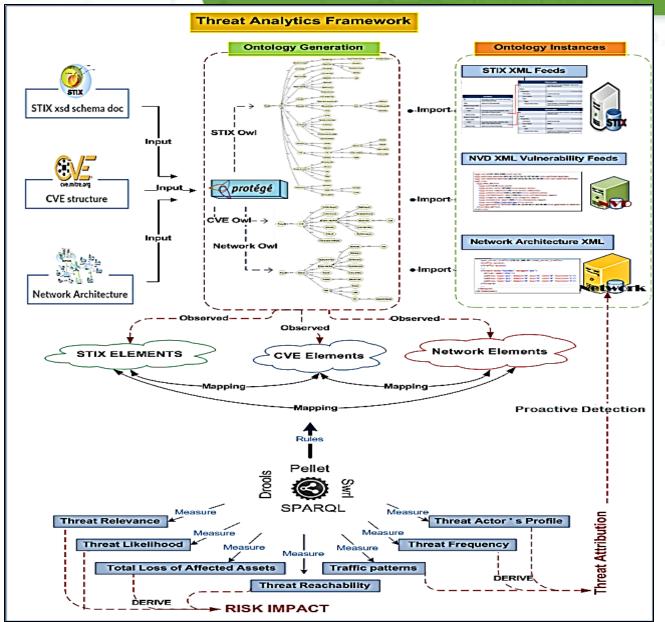


#### Industrial and Business Relevance

- Proactive security: Ensuring that the current implementation of the network reserves its mission in the face of cyber threats.
- Automatic: Automatic transition from "threat intelligence" to "mitigation actions".
- Cost-effective: fixes of critical vulnerabilities and risky configurations are based on cost, usability and security requirements.



## STIXChecker Process Flow





# LOGICAL FORMALIZATION OF STIX (OWL/SMT)



### **Ontology**

- Leveraged
  - Existing work from Vistology
  - NVD Database
- Domains and Restrictions

 $Indicator \equiv STIX \cup$ 

 $\in hasObservable\ has\ Observables \cup$ 

 $\exists \ hasIndicators \ some \ Indicators \cup$ 

 $\exists \ hasRelatedTTP \ some \ TTP$ 

 $Host \equiv Network \cup$ 

 $\exists \ hasFirewall \ some \ Firewall \cup$ 

- $> hasConnectedHost some Host \cup$
- > hasRouter min 1∪
- $\geq hasHostName\ min\ 1 \cup$
- $\geq hasHostIP min 1 \cup$

 $\exists \ has Vulnerable Software \ some \ Installled S$ 

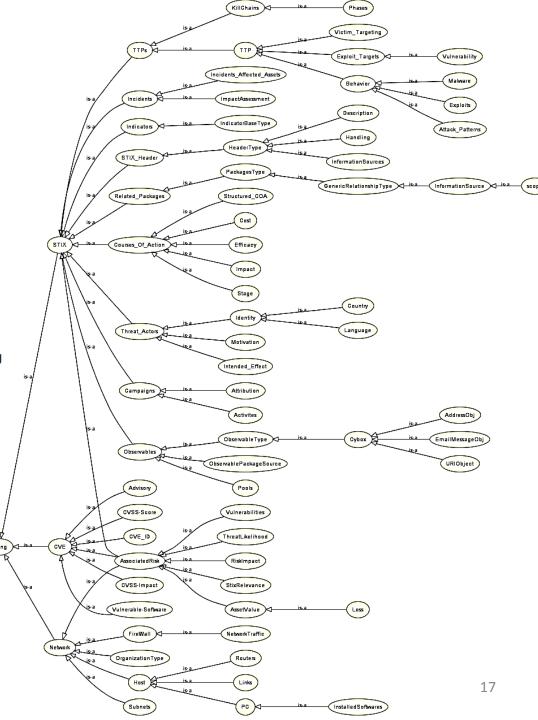
 $CVE \equiv owl : Thing \cup$ 

 $\leq hasCVE\_ID\ max\ 1 \cup$ 

 $\leq hasCVSS\_baseScore\ max\ 1$ 

 $\geq hasVulnerableSoftware min 1 \cup$ 

 $\geq hasAdvisory min 1$ 





# DETERMINING NETWORK RELEVANCE



#### **Relevance Factors and Associated Weights**

## Relevance Scoring

$$S_{i} = \frac{E_{s} \cap E_{n}}{\bar{E}}$$

#### where:

 $E_s$  is set of relevance elements found in STIX  $E_n$  is set of relevance elements received in network  $\bar{E}$  is set of all available relevance elements

#### Threat Likelihood

$$L = \max_{0 \le S_i \le 1} \frac{\sum_{i=0}^{N} S_i \times W_i}{\sum_{i=0}^{N} \bar{S}_i \times W_i}$$

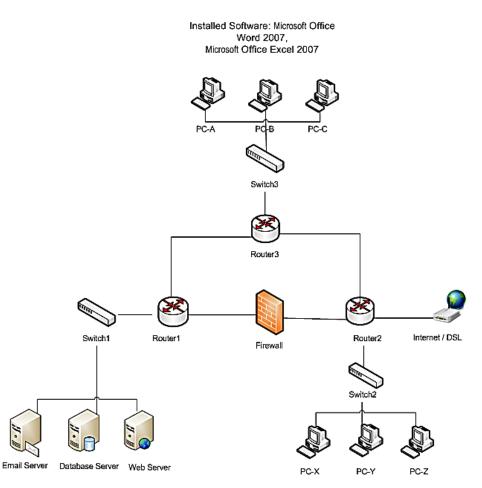
#### where:

N is number of relevance factors F  $S_i$  is received relevance score for  $F_i$   $\bar{S}_i$  is maximum relevance score for  $F_i$   $W_i$  is assigned weight for  $F_i$ 

F	$W_i$
hasCVE_Relevance	5
hasAssetsRelevance	4
hasCIA_Relevance	4
hasTargetedLocationRelevance	3
hasMotivationRelevance	2
hasOrganizationRelevance	1
hasImpactRelevance	1
hasTargetedLanguageRelevance	1
hasSecurityCompromiseRelevance	1



## Example Case study—Red October APT & Ashley Madisor



Installed Software: Microsoft Office Word 2003

	Red October	Network Ele-	Relevance
	STIX Elemenets	ments $(E_n)$	$Score(S_i)$
	$(E_s)$		
hasCVE_Relevance	CVE-2012-0158,	CVE-2009-	1
	CVE-2010-3333,	3129, CVE-	
	CVE-2009-3129	2010-3333,	
		CVE-2012-	
		0158	
hasAssetsRelevance	Servers, Routers,	Servers,	0.66
	Switches,	Routers,	
	Persons, PCs	Switches and	
	and Mobile	PCs	
	Phones		
hasTargetedLocation Rel-	USA	USA	0.25
evance			
hasTargetedLanguage	English	English	1
Relevance			
hasCIA_Relevance	Confidentiality,	Confidentiality,	1
	Integrity,	Integrity,	
	Availability	Availability	

#### **Network-STIX Relevance**

F	$S_i \times W_i$	$ar{S}  imes oldsymbol{W_i}$	
hasCVE_Relevance	5	5	
hasAssetsRelevance	2.64	4	
hasCIA_Relevance	4	4	
hasTargetedLocationRelevance	0.75	3	
hasMotivationRelevance	0	2	
hasOrganizationRelevance	0	1	
hasImpactRelevance	0	1	
hasTargetedLanguageRelevance	1	1	
hasSecurityCompromiseRelevance	0	1	
SUM	13.39	22	
L		13.39/22	=

Threat Likelihood

0.60



# IMPACT ANALYSIS & KILLING THE CYBER KILL-CHAIN



#### STIX threat Modeling

- STIX feeds can be generically formalized as steps within a kill chain.
- A phase can be decomposed into one or multiple TTPs

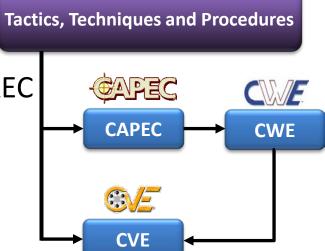
Kill chain-phase= 
$$\forall_i^{p-1} TTP_i \longrightarrow TTP_{i+1}$$

- TTPs can further be broken up into:
  - Attack Patterns given through CAPEC
  - Exploit Targets given through CVE
  - Malwares behavior using MAEC



CAPEC can be broken into CWEs

CAPEC = 
$$\bigvee_{i}^{n} CWE_{i}$$





#### **Proposed Impact Metric**

 Measures the damage inflicted by STIX threats and the contribution of each kill chain phase to the total damage.

$$\mathsf{Impact}\,(\mathsf{d},\mathcal{S}) \,=\, \begin{cases} V_d \, *A_d * \mathsf{progress} \\ V_d \, *A_d * \mathsf{progress} + (w * \sum_{i \in \mathbb{F}} Impact(i,\mathcal{S}-1)) \end{cases} \qquad \qquad \mathcal{S} = 0 \text{ or } \mathbb{F} = 0$$
 
$$\mathcal{S} > 0 \text{ or } \mathbb{F} > 0$$

#### Where

 $V_d$ : vulnerability score of the host (likelihood \*severity).

 $A_d$ : the asset value of the host d.

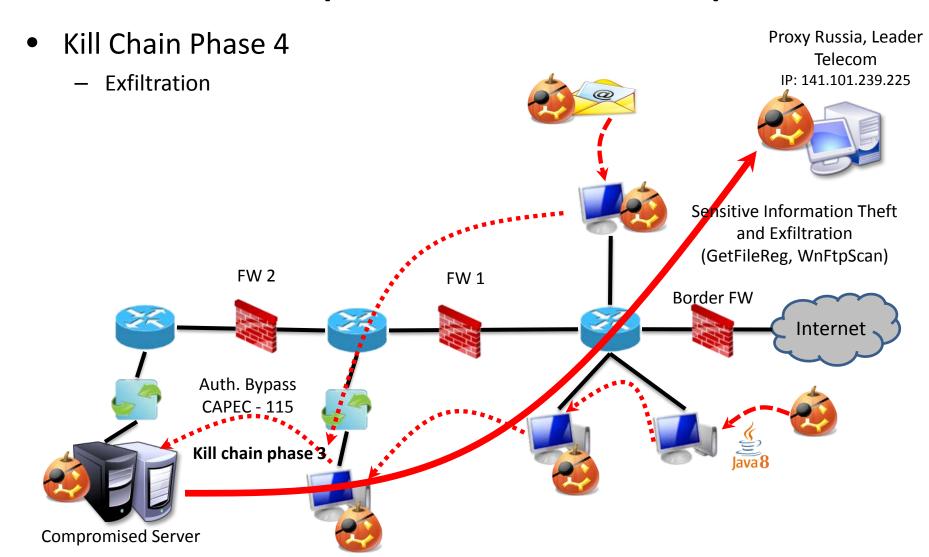
 $\mathbf{r}$ : set of the reachable hosts which are vulnerable to the next **kill chain phase.** (i.e., hosts that have vulnerabilities which enable the next phase).

S: the number of maximum recursion steps (recursion threshold) S = 0 means a leaf node.

W: a weight variable (keeps getting smaller with every recursive call due to indirect damage) progress: the percentage of completed attack phases upon successfully compromising the selected host.



#### **Network Impact – Red October Example**

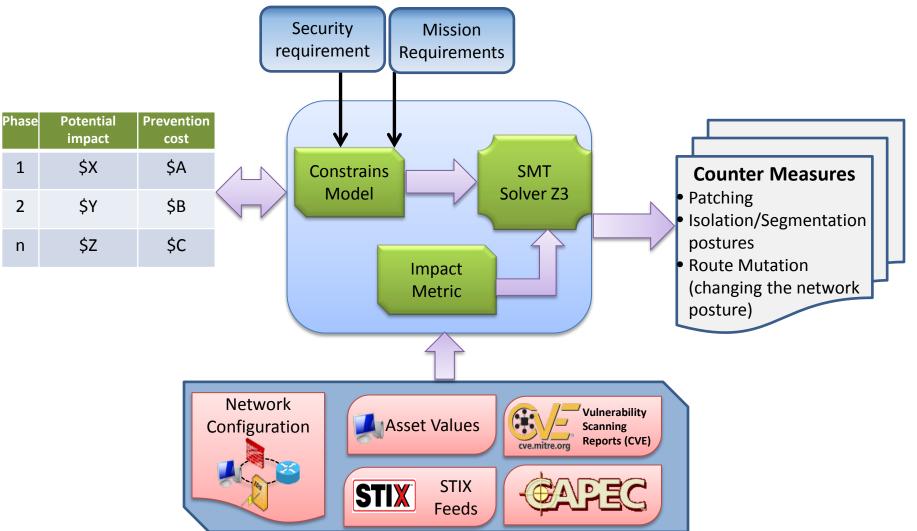




## FROM STIX TO ACTIONS



#### Inside the Reasoning Engine -- From STIX To Actions





# DATA-DRIVEN VISUAL ANALYTICS: REPUTATION ... ETC



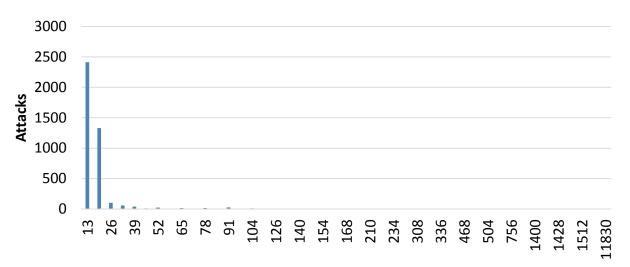
#### Visualization: Initial Exploratory Experiments

- Initial Dataset
  - 2 categories polled from Hailataxii
    - MalwareDomainList
    - CyberCrime\_Tracker
  - 10-minute time window
    - 2015-06-25T13:00 2015-06-25T13:10
  - 158,510 STIX Documents retrieved
  - -~ 482 MB Size



#### Problem: Dealing with a Huge Dataset

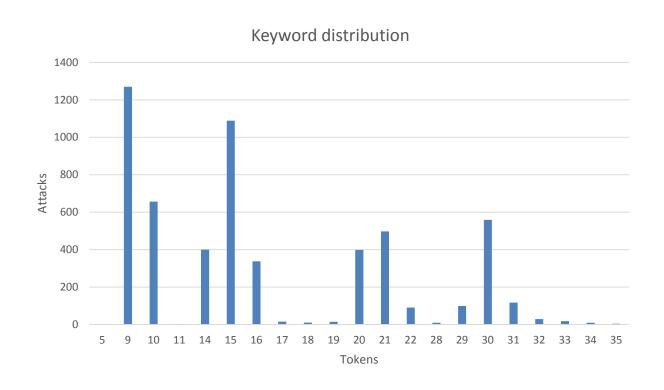
- Experiment 1: Grouping of 158,510 documents
  - Considered XML structures
    - 6 different groups
  - Considered content
    - 5,623 different groups (attacks)
      - 96.5 % reduction
    - Each group represents an attack
      - Avg docs in each group: 28





### Problem: Dealing with a Huge Dataset (Continued)

- Experiment 2: Richness of attack information
  - Avg: 16.3 words
  - Excluding: stop words



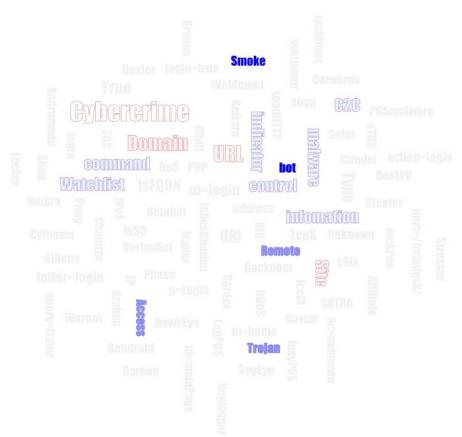


- Word Cloud (created from contents)
  - To explore the content of the dataset and to see what is relevant to INW
  - http://cyberdna.uncc.edu/inw/stix/words.php





- Keyword correlation
  - Intensity of blue => Stronger relationship
  - Gray without border => Not a related word





Urls related to each keyword (co-occurrence)





#### Related STIX Documents

- User-friendly sampling of files for each attack (group)
- Number of STIX documents in each attack
- Attack Duration

#### **PONY**

Attack ID	Sample File	Start Date	End Date	Count
13	guest.CyberCrime Tracker STIX111 12015 06 25T17 00 00 889585 00 00.xml	2015-06-25 17:00:00	2015-06-25 17:09:38	13
14	quest.CyberCrime Tracker STIX111 t2015 06 25T17 00 00 916096 00 00.xml	2015-06-25 17:00:00	2015-06-25 17:09:40	494
22	guest-CyberCrime Tracker STIX111 12015 06 25T17 00 00 915034 00 00.xml	2015-06-25 17:00:00	2015-06-25 17:09:38	13
62	quest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 019254 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
65	quest.CyberCrime Tracker STIX111 12015 06 25T17 00 01 026258 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
71	quest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 040738 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
77	quest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 056161 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
91	quest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 130903 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
93	guest.CyberCrime Tracker STIX111 12015 06 25T17 00 01 137958 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
96	guest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 144995 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
104	guest.CyberCrime_Tracker_STIX111_t2015_06_25T17_00_01_166655_00_00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
106	guest.CyberCrime Tracker STIX111 t2015 06 25T17 00 01 173758 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
108	guest CyberCrime Tracker STIX111 12015 06 25T17 00 01 180305 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:38	13
128	guest CyberCrime Tracker STIX111 12015 06 25T17 00 01 306155 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:43	39
144	guest CyberCrime Tracker STIX111 12015 06 25T17 00 01 363733 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:43	26
208	guest CyberCrime Tracker STIX111 12015 06 25T17 00 01 523118 00 00.xml	2015-06-25 17:00:01	2015-06-25 17:09:44	26
222	TO COLOR TO THE CONTRACT OF STREET OF STREET OF STREET	2015 20 25 17 20 21	2015 22 25 17 22 15	



#### **Initial Exploratory Experiments**

- Initial Dataset
  - -1 Channels in Hailataxii
    - CyberCrime\_Tracker
  - -8-hour time window
    - 2015-06-25T12:00 2015-06-25T19:00
  - -3,476,792 STIX Documents
  - -About ~10GB



#### **Initial Exploratory Experiments**

- Grouping
  - 6 different templates
  - Repeated chunks of text
    - Descriptions
      - [This domain <domain\_name> has been identified as a command and control site for <malware\_name> malware by cybercrime-tracker.net. For more detailed infomation about this indicator go to [CAUTION!!Read-URL-Before-Click] [http://cybercrimetracker.net/index.php].</indicator:Description>]
    - Term\_of\_use
    - Statement



## **Initial Exploratory Experiments**

- Visualization of Related domains
  - https://public.tableau.com/profile/hu4869#!/vizhome/stix/Sheet2





## Treemap view Identifies which attacks are Important



<sup>\*</sup> each color represents an attack



# OBJECTIVE REPUTATION OF CYBER THREAT INTELLIGENCE SOURCES—

TOWARD PURIFICATION AND CLASSIFICATION



#### **Motivation and Goals**

- STIX will include noisy and possible malicious sources
- How do you know which CTI sources to consider:
  - Removing noise: duplication, bogus etc
  - Priority-based classification
- Creating community self-awareness and accountability consuming
- Allow customers to narrow their search and act faster
- Proposed Ranking Service is based on:
  - 1. Threat-source profiling based on time-series and information theatric analysis
  - 2. Multi-Source correlation using clustering and visualization for STIX inter-relationship and source inter-dependency analysis
  - 3. Sentiment Analysis and Consumer Reports
  - 4. Integrating **Cyber Intelligence** information to enrich the reputation analysis



Time-



### Key Features in selection of Reputable CTI sources

- Number of entries (signal/noise)
- Certainty (blind aggregation, lack of context)
- Type of badness (only certain types e.g. C&C)
- Standards followed (direct input to network FW?)
- Update Frequency (daily, hourly, real-time)
- Varying level of detail
- Frequency of false positives
- Threat Querying by application and features

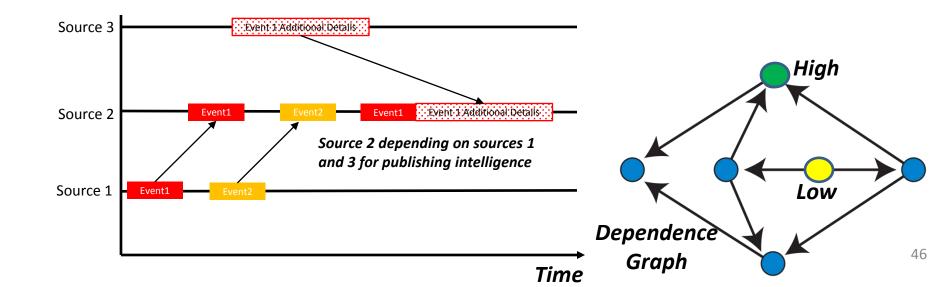
Recommended Sources  Exclusive Ratings Rating of over xx Sources				
	Coverage	Dependency	Standards	Trustworthiness
Source 1				
Source 2				

**Consumer Reports Best & Worst CTI Sources 2015** 



## Development of scores/metrics

- Detect faster: How much will feed reduce time to detect?
- Detect Better: How much will feed enable me to detect what I would otherwise miss?
- Dependency Score: for decision making about independence of source





### **Value Proposition**

- Ability to rank source reputation and purchase source based on:
  - The specialization of the threat source
  - Quantitative/qualitative scoring
  - Feature wish-list search
  - Partially ordered (ranked) lists
  - Coverage
  - Suggestions based on user requirements
    - Single best source of threat intelligence
    - Customization of services from multiple sources



# PRELIMINARY RESULTS

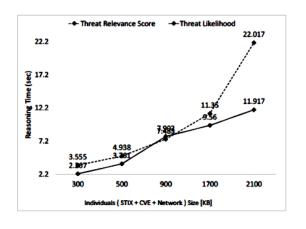


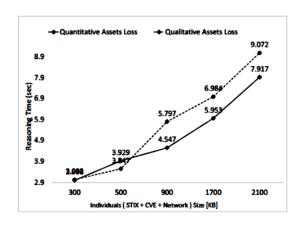
## Validation Experiments of our Preliminary Results

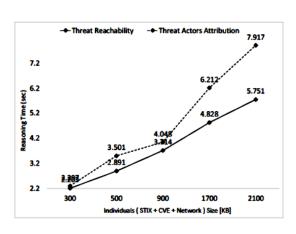
- We used 20 different case studies of various attack represented in STIX
  - X are already made ones including Red October
     APT attack
  - Y are created by our team based on CTI sources such as ThreatConnect including Ashley Madison attack
- Validation Methodology



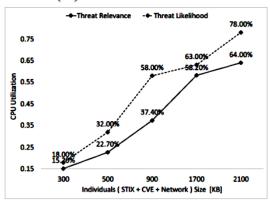
#### **Reasoning Time and CPU Utilization**



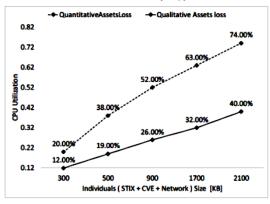




(a) Relevance Score  $(S_i)$  & Threat Likelihood (L)

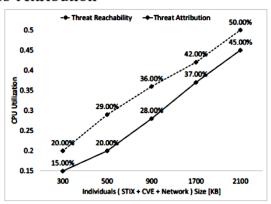


(b) Quantitative Assets Loss  $(A_n)$  & Qualitative Assets Loss  $(A_l)$ 



(b) Quantitative Assets Loss  $(A_n)$  & Qualitative Assets Loss  $(A_l)$ 

(c) Threat Reachability (R) & Threat Actors Attribution

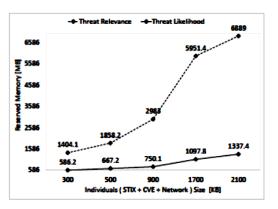


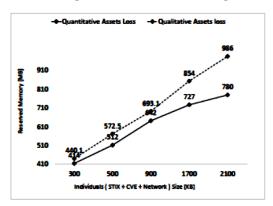
Loss  $(A_n)$  & (c) Threat Reachability (R) & Threat Actors Attribution

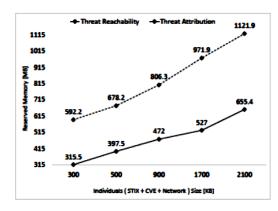
(a) Relevance Score  $(S_i)$  & Threat Likelihood (L)



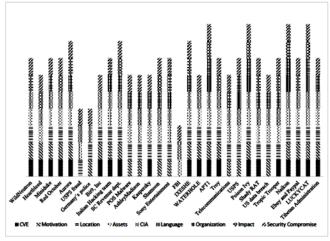
## **Memory Consumption**

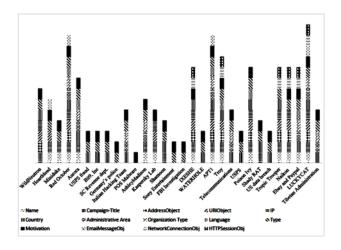






(a) Relevance Score  $(S_i)$  & Threat Likeli-(b) Quantitative Assets Loss  $(A_n)$  & Qual-(c) Threat Reachability (R) & Threat Achood (L) itative Assets loss  $(A_l)$  tors Attribution





51

Fig. 10: Relevance Factors (F) found in STIX

Fig. 11: Threat Actor's Attributes found in STIX



#### Conclusion

- STIX Threat Information Sharing is the right step in the right direction for cybersecurity automation
- But many others steps have to follow to create incentive (usability and effectiveness) of STIX-based CTI.
- Our experience shows both formal- and data-driven approaches to address critical challenges and bridge this gab between CTI sharing and usability/effectiveness
- This is the tip of the iceberg: More research and development is needed in this direction ...
- Relevance: Invitation visit and join the NSF Center on [Security] Configuration Analytics and Automation (www.ccaa-nsf.org);



## Questions

