

Exploiting the dark triad for national defense capabilities



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What is all about?

The problem:

An insider as a severe threat to national defense

The theory:

Narcissists tend to turn insiders

The defense:

Reveal narcissists

The data source:

Online Social Networks and the Web 2.0

The tool:

NEREUS Framework



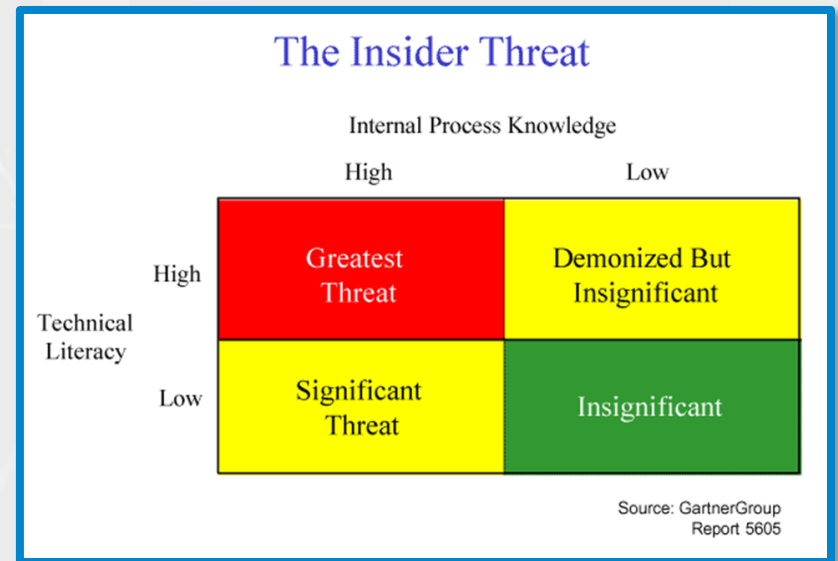
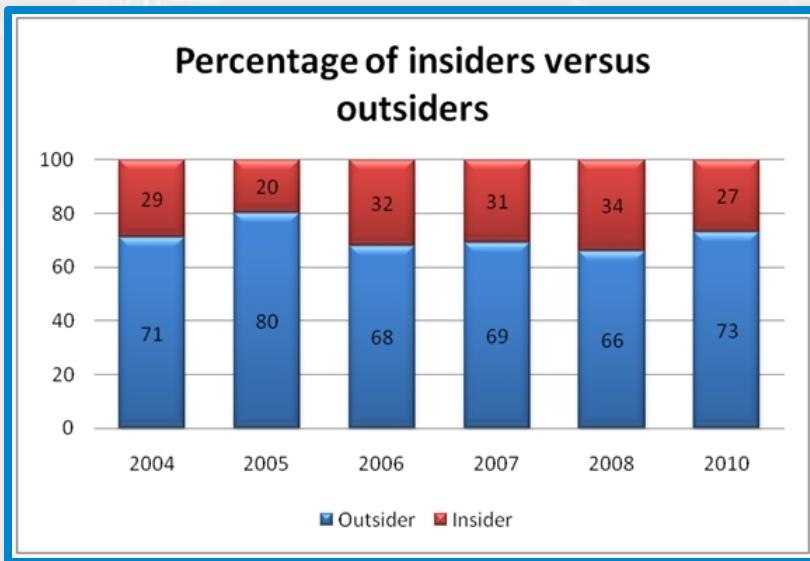
The problem: The Insider threat

Insider threat was the **top priority to protect** for 2014

Insiders are persons who:

- are legitimately given access rights to a Critical Infrastructure
- misuse their privileges and violate security policy

Insiders consist the **top source** of data breaches



Source:

CERT, *Cyber Security Watch Survey: How Bad Is the Threat?*, Carnegie Mellon University, USA, 2014.

The theory: The Dark Triad

- **Dark Triad personality traits**
 - Narcissism: Inflated self-view and focus on themselves
 - Machiavellianism: Manipulative personality
 - Psychopathy: High impulsivity and thrill-seeking, along with low empathy and low anxiety
- Dark Triad traits are used to extend Five Factor Model (FFM) to represent **socially malevolent behavior**
- The wealth of data provided by OSN users has opened the door to a new way of **analyzing personality**
- Ability to **exploit narcissism** personality trait
 - Narcissism trait is also examined by Shaw and the FBI

Sources:

M. Maasberg, J. Warren, N. Beebe, "The Dark Side of the Insider: Detecting the Insider Threat through examination of Dark Triad personality traits".

Federal Bureau of Investigation, *The insider threat: An introduction to detecting and deterring an insider spy*, USA, 2012.

E. Shaw, K. Ruby, J. Post, "The insider threat to information systems: The psychology of the dangerous insider", *Security Awareness Bulletin*, vol. 2, no. 98, pp. 1--10. 1998.

The defense: Open Source Intelligence

Open Source Intelligence (OSINT) is produced from publicly available information, which is:

- Collected, exploited and disseminated in a **timely** manner
- Offered to an **appropriate** audience
- Used for the purpose of addressing a specific **intelligence requirement**

Publicly available information refers to (not only):

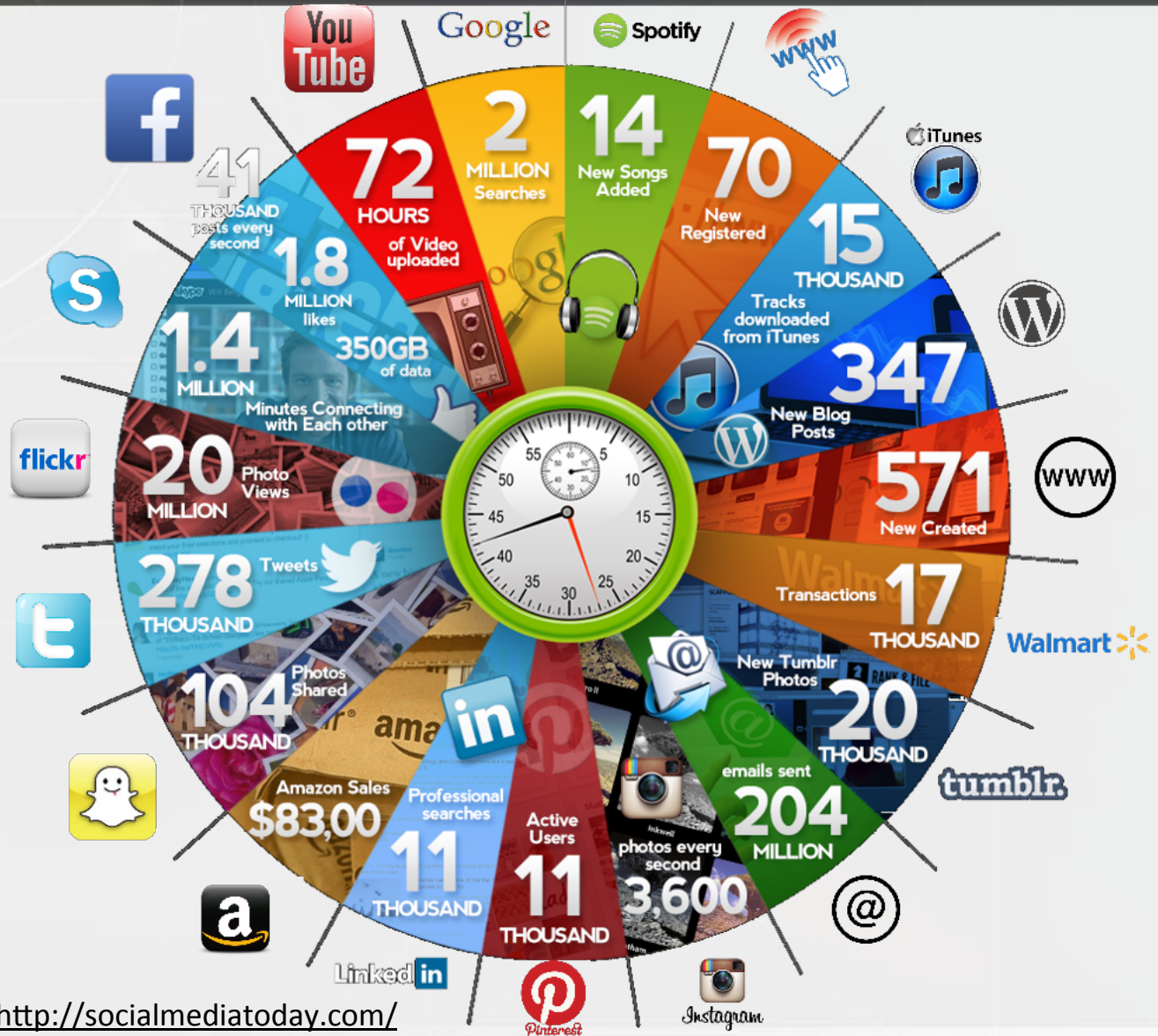
- Traditional media (e.g. television, newspapers, radio, magazines, etc.)
- Web-based communities (e.g. social networking sites, blogs, etc.)
- Public data (e.g. government reports, official data, public hearings, etc.)
- Amateur observation/reporting (e.g. amateur spotters, radio monitors, etc.)

OSINT defined by US Dept. of Defense (Public Law 109-163, Sec. 931, National Defense Authorization Act for Fiscal Year 2006).

SOCMINT is produced from Online Social Networks & the Web 2.0



The data source: Web 2.0 & Online Social Networks



Source: <http://socialmediatoday.com/>

The tool: The **NEREUS** Framework

NEREUS Framework

OSN: Twitter



Tools used for the open data analysis

Science

Theory

Informatics

Graph Theory

Content Analysis

Sociology
Psychology

Theory of Planned Behavior

Social Learning Theory

Application: Insider threat detection/prediction, influential users detection, means of communication evaluation, etc.



The NEREUS Framework in a nutshell



Predicting & identifying potential insiders



Researchers' compliance with ethical standards

YES



Legal Expert

YES

Critical infrastructures
National security
Public interest



Twitter Users

Content generation



Twitter

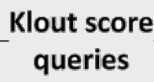
Crawling & storing



Our crawling server



Klout score server



Klout score queries



Klout score api collector

Legend		
Web 2.0 Medium:	Twitter	
Domain Expert:	Psychologist	



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Category	Influence valuation	Klout score	Usage valuation
Loners	0 - 90	3.55 - 11.07	0 - 500
Individuals	90 - 283	11.07 - 26.0	500 - 4.500
Known users	283 - 1.011	26.0 - 50.0	4.500 - 21.000
Mass Media & Personas	1.011 - 3.604	50.0 - 81.99	21.000 - 56.9000

Content Aggregator

Usage intensity valuation

Indegree/oudegree aggregator

Influence valuation

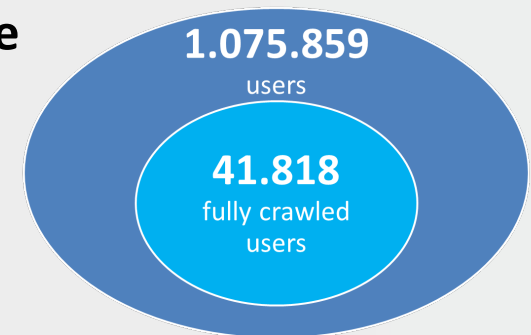
User classification according to categories

The dataset utilized



- Focus on a Greek **Twitter** community:
 - Context sensitive research
 - Utilize ethnological features rooted in locality
 - Extract and analyze results
- Analysis of **content** and measures of **user influence** and **usage intensity**
- User Categories: Follower, Following, Retweeter
- Graph:
 - Each user is a node
 - Every interaction is a directed edge
- **41.818** fully crawled users (personal & statistical data)
 - Name, ID, personal description, URL, language, geolocation, profile state, lists, # of following/followers, tweets, # of favorites, # of mentions, # of retweets

Twitter (Greece, 2012-13)



7.125.561 connections
among them





Graph theoretical and content analysis

Strongly connected components:

- There exists 1 large component (153.121 nodes connected to each other) and several smaller ones

Node Loneliness:

- 99% of users connected to someone

Small World phenomenon:

- Every user lies <6 hops away from anyone else

Indegree Distribution:

- # of users following each user
- Average 13.2 followers/user

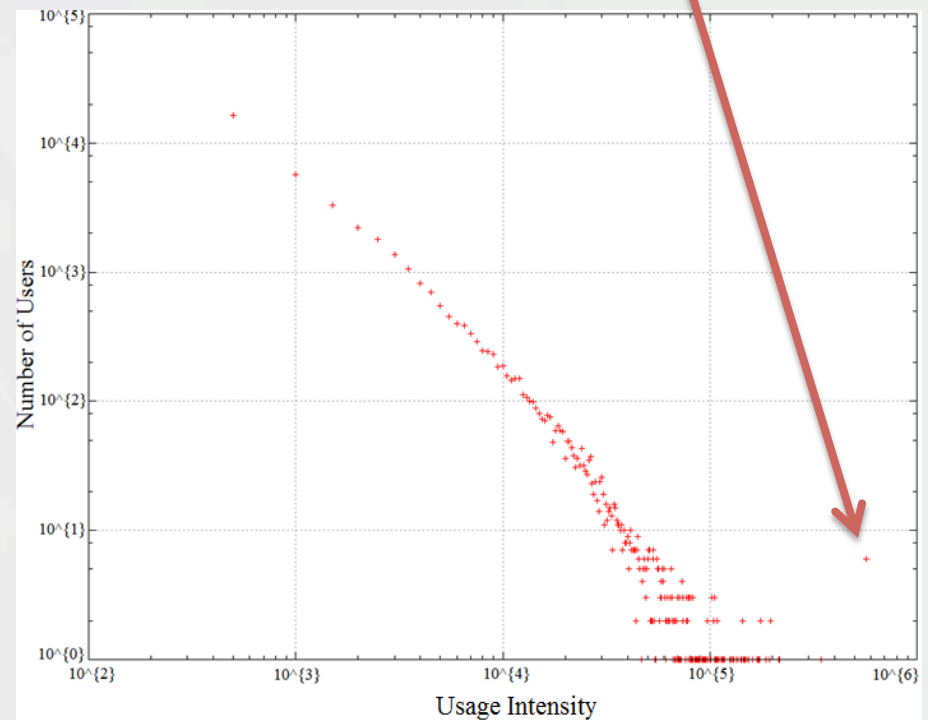
Outdegree Distribution:

- # of users each user follows
- Average 11 followers/user

Usage Intensity Distribution:

Weighted aggregation of: {# of followers, # of followings, tweets, retweets, mentions, favorites, lists}

Important cluster of users



Category	Influence valuation	Klout score	Usage valuation
Loners	0 - 90	3.55 - 11.07	0 - 500
Individuals	90 - 283	11.07 - 26.0	500 - 4.500
Known users	283 - 1.011	26.0 - 50.0	4.500 - 21.000
Mass Media & Personas	1.011 - 3.604	50.0 - 81.99	21.000 - 56.900

Conclusions

- ✓ Web 2.0 produces vast amounts of **crawable** information and OSINT can transform it into **intelligence**.
- ✓ OSINT can assist in detecting the Dark Triad traits (**narcissistic behavior**, etc.).
- ✓ OSINT can help in **predicting insiders**, in **predicting delinquent behavior**, and in **enhancing national defense**.
- ✓ OSINT **intrusive nature** dictates use for **specific** purposes, according to law.



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